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## A Monte Carlo approach for the fully probabilistic evaluation of operability in ship dynamic positioning scenarios *⊙*

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### A Monte Carlo Approach for the Fully Probabilistic Evaluation of Operability in Ship Dynamic Positioning Scenarios

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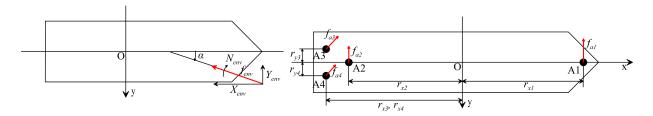
**Abstract.** The Dynamic Positioning system allows a vessel to keep a precise position and heading during stationing operations in a rough sea by using onboard actuators only. During the design phase, it is mandatory to identify the capability of the system actuators to counteract the environmental forces. Conventional predictions are limited to the estimation of a maximum sustainable wind speed on predefined encounter angles by estimating the corresponding wave parameters with questionable standard deterministic correlations. The proposed approach aims at determining the dynamic positioning performances by using site-specific long-term environmental conditions which are modelled with joint distributions of wind and wave parameters. To this end, the operability of the dynamic positioning system is evaluated as a non-deterministic multidimensional Monte Carlo integration process, based on the sampling of environmental joint distributions. For each environmental condition, a quasi-static dynamic positioning analysis is performed solving the equilibrium between external forces and the vessel's actuators through a non-linear thrust allocation algorithm. The proposed methodology is applied to a reference offshore ship in five different operative geographic areas, highlighting the suitability of the calculation methodology for site-specific operability predictions.

#### INTRODUCTION

The Dynamic Positioning (DP) characteristics determination for a ship requires the knowledge of reference or site-specific environmental conditions. During the preliminary design phase, the description of the DP working environment of the vessel is not always correctly considered, as the methods usually employed to perform DP calculations only allow for the usage of fixed prescribed environmental conditions [1, 2, 3]. Such conditions are just deterministic wind-wave correlations, providing wave parameters (significant wave height  $H_s$  and peak period  $T_p$ ) and wind speed  $V_w$  specific for reference DP predictions. These correlations allow for determining of a maximum wind speed envelope bounding the wind speed domain at each heading angle  $\chi$  where the DP system holds the position [4]. However, there are different environmental characteristics descriptions for specific sea areas, giving a more detailed characterisation between the occurrence and correlations of wind and wave parameters [5]. Such models are potentially available since the preliminary phases of design but require the adoption of alternative calculation processes to assess DP system performances.

An alternative to conventional DP predictions is a quasi-probabilistic method developed to handle the joint occurrence of wave parameters  $H_s$  and  $T_p$ . This methodology, called the *scatter diagram approach* [6], considers the couples  $(H_s, T_p)$  present in the long-term statistic of a sea area (the scatter diagram) and uses a deterministic procedure to evaluate the wind speed  $V_w$  from  $H_s$  and  $T_p$  occurrence. Therefore, the DP calculations refer to the cells of a scatter diagram. As the main relevant output, the *scatter diagram approach* does not produce a limiting wind speed envelope but calculates the DP system operability in a given sea area. This is a shortcut for the designers.

Nonetheless, it is possible to reproduce the long-term environmental conditions of a given sea area also considering the joint occurrence of wind and waves [7], thus employing a trivariate joint distribution of  $V_w$ ,  $H_s$  and  $T_p$ . A specific and fully probabilistic description of the environmental parameters does not allow evaluating DP system operability through calculations on specific predetermined conditions. Here, a novel approach calculates the operability as numerical integration, implementing a Monte Carlo (MC) process for DP performance estimation. The implemented procedure samples the environmental conditions from the joint distribution, which are the inputs for DP calculations. In this study, the DP analysis follows a quasi-static approach. That means solving the equilibrium between external forces and vessel actuators through a thrust allocation algorithm. The proposed MC procedure uses a non-linear thrust allocation algorithm [8] to solve the quasi-static problem. The paper reports the application of the MC process to a reference vessel for different operative sea areas of interest. The study analyses the integration convergence and the



**FIGURE 1.** Reference system for quasi-static DP analysis.

FIGURE 2. Actuators layout for the reference OSV.

TABLE I. Reference OSV main characteristics.

Name	Symbol		Unit
Length between perpendiculars	$L_{PP}$	72.00	m
Length overall	$L_{OA}$	78.35	m
Maximum breadth	B	16.00	m
Operative draught	T	4.05	m
Volume	$\nabla$	3245.21	$m^3$
Lateral exposed wind area	$A_L$	854.10	$m^2$
Transversal exposed wind area	$A_T$	187.40	m <sup>2</sup>

**TABLE II.** Reference OSV actuators configuration.

Actuator ID and type	$r_{\chi}$ (m)	<i>r</i> <sub>y</sub> (m)	f <sub>amax</sub> (kN)
A1 (tunnel thruster)	32.00	0.00	73.50
A2 (tunnel thruster)	-30.00	0.00	62.50
A3 (azimuth thruster)	-36.00	-3.00	240.00
A4 (azimuth thruster)	-36.00	3.00	240.00

variance of the results introduced by the MC process on the DP operability. The results allow identifying a suitable confidence interval for the operability determination, including the heading angle  $\chi$  in the sampling process.

#### **QUASI-STATIC DP ANALYSIS**

Two different methods allow for evaluating the stationing capability of offshore ships: quasi-static analysis [9, 10] and time-domain simulations [11, 12]. The latter method is more time-consuming and requires knowledge of the control parameters of the effective DP system installed onboard the analysed ship [13]. Therefore, time-domain simulations are particularly suitable for the final assessment of DP performances for the definitive system mounted onboard. On the other hand, quasi-static calculations are the preferred option when all the ship and DP system parameters are not fixed or known, but an assessment of DP performances is needed. For such a reason, quasi-static DP analyses are a standard praxis in the design of offshore ships.

The so-called quasi-static analysis of the DP performances consists of the resolution of the static equilibrium of forces/moment acting on the ship's horizontal plane, which means primarily environmental and actuator loads taking into account dynamic effects with dynamic allowances on input loads. Considering, for instance, a body-fixed reference system centered at midship (see Figure 1), the static equilibrium in 3 degrees of freedom has the following form:

$$f_{env} = A(\alpha) f_a \tag{1}$$

where  $f_{env} = [X_{env}, Y_{env}, N_{env}]^{\mathrm{T}}$  is the environmental loads vector,  $f_{\alpha} = [f_{a_1}, \cdots, f_{a_{N_a}}]^{\mathrm{T}}$  is the vector of the thrust delivered by the  $N_a$  actuators and  $A(\alpha) \in \mathbb{R}^{3 \times N_a}$  is a matrix containing the actuator positions and the thrust orientations vector  $\alpha = [\alpha_1, \cdots, \alpha_{N_a}]^{\mathrm{T}}$ . The definition of  $A(\alpha)$  is vessel-specific and depends on the type of actuators installed onboard. Considering  $A(\alpha)$  as a set of  $N_a$  column vectors  $a_i \in \mathbb{R}^3$ , the following forms apply to the actuators commonly used for DP purposes:

$$\boldsymbol{a}_{i} = \begin{cases} \left[1, 0, -r_{y_{i}}\right]^{\mathrm{T}} & \text{for main propellers} \\ \left[0, 1, r_{x_{i}}\right]^{\mathrm{T}} & \text{for fixed tunnel thusters or rudders} \\ \left[\cos \alpha_{i}, \sin \alpha_{i}, r_{x_{i}} \sin \alpha_{i} - r_{y_{i}} \cos \alpha_{i}\right]^{\mathrm{T}} & \text{for azimuth thusters} \end{cases}$$

$$(2)$$

where  $r_{x_i}$  and  $r_{y_i}$  are the transversal and longitudinal positions of the onboard actuators.

The present study uses as a reference an offshore supply vessel (OSV) having the main dimensions and actuators configuration reported in Table I and Table II, respectively. The actuator layout is also shown in Figure 2. The reference OSV has two steerable and two fixed tunnel thrusters as actuators. Therefore, remembering actuators definition given in Eq. (2), matrix  $\mathbf{A}(\alpha) \in \mathbb{R}^{3\times 4}$  assumes the following form:

$$\mathbf{A}(\alpha) = \begin{bmatrix} 0 & 0 & \cos \alpha_3 & \cos \alpha_4 \\ 1 & 1 & \sin \alpha_3 & \sin \alpha_4 \\ r_{x_1} & r_{x_2} & r_{x_3} \sin \alpha_3 - r_{y_3} \cos \alpha_3 & r_{x_4} \sin \alpha_4 - r_{y_4} \cos \alpha_4 \end{bmatrix}$$
(3)

Then, the static equilibrium for the reference OSV results by substituting Eq. (3) in Eq. (1):

$$\begin{bmatrix} X_{env} \\ Y_{env} \\ N_{env} \end{bmatrix} = \begin{bmatrix} 0 & 0 & \cos \alpha_3 & \cos \alpha_4 \\ 1 & 1 & \sin \alpha_3 & \sin \alpha_4 \\ r_{x_1} & r_{x_2} & r_{x_3} \sin \alpha_3 - r_{y_3} \cos \alpha_3 & r_{x_4} \sin \alpha_4 - r_{y_4} \cos \alpha_4 \end{bmatrix} \begin{bmatrix} f_{a_1} \\ f_{a_2} \\ f_{a_3} \\ f_{a_4} \end{bmatrix}$$
(4)

The equilibrium system in Eq.4 has a number of unknowns  $N_u$ =6. The unknowns can be grouped in a reference vector  $\boldsymbol{u} = [\boldsymbol{f_a}, \boldsymbol{\alpha}]^{\mathrm{T}} = [f_{a_1}, f_{a_2}, f_{a_3}, f_{a_4}, \alpha_3, \alpha_4]^{\mathrm{T}} \in \mathbb{R}^6$ . As general rule, an azimuth thruster has 2 unknowns while tunnel thrusters, rudders and fixed propeller have one unknown each. Therefore,  $N_u$  can be determined for all vessel types and any actuators configuration as  $N_u = N_a + N_{at}$ , where  $N_{at}$  is the number of azimuth thrusters installed onboard. Hence, for quasi-static DP calculations, the resolution of Eq.(4) admits infinite solutions when  $N_u > 3$ . That means an unique solution of the equilibrium is admitted only in the case of unrealistic DP actuator layouts as a combination of 1 azimuth and 1 tunnel thrusters or the presence of 3 tunnel thrusters only. For such reasons, alternative solutions are needed to solve the equilibrium system.

#### **Thrust allocation**

The resolution of the static equilibrium system in Eq.(4) imply the implementation of a suitable algorithm to determine u, means the proper distribution  $f_a$  and orientation  $\alpha$  of the thrust among the  $N_a$  actuators. The computation of u is adaptable to a constrained optimisation problem, minimising an approximated expression of the absorbed power as a function of  $f_a$ :

$$\min(z) = \sum_{i=1}^{N_a} |f_{a_i}|^{3/2} \tag{5}$$

subject to the following constraints:

$$A(\alpha) f_{\alpha} = f_{env} \tag{6a}$$

$$f_{a_{min}} \le f_{a} \le f_{a_{max}} \tag{6b}$$

$$\alpha_{min} \le \alpha \le \alpha_{max}$$
 (6c)

The constraints described by Eq.(6a) represents the static equilibrium system of Eq.(4), Eq.(6b) is used to allocate thrust between the actuator saturation limits, and Eq.(6c) sets the feasible sectors of azimuth angles  $\alpha_i$ .

The optimisation problem provided by Eq.(5) and Eq.(6) is nonconvex and nonlinear and can be solved by using a nonlinear iterative optimisation algorithm [8]. Such a method, can handle the thrust allocation problem considering as unknowns  $f_a$  and  $\alpha$ , or considering the thrust components  $f_{ax} = f_a \cos \alpha$  and  $f_{ay} = f_a \sin \alpha$ , expressed in the body fixed reference system [14]. With this change of variables, the objective function remains nonlinear but Eq.(6a) becomes linear. The constraints of Eq.(6b) become nonlinear but can be easily linearised to reduce the resolution complexity without losing accuracy in the final result. The implementation of azimuth constraints in Eq.(6c) is substituted by automatically adding equality constraints when needed, forcing the actuator to stay in the feasible region.

As a results the thrust allocation algorithm determines, beside the specific thrust intensity and orientation of the actuators, whether the DP system is capable or not to keep the position, thus respecting the equilibrium of Eq.(1).

#### **Environmental loads**

The environmental loads  $f_{env}$  acting on the ship are the main external forces influencing the equilibrium needed to perform a quasi-static DP analysis. It is then necessary to analyse more in detail their evaluation. The conventional breakdown of the environmental forces is as follows:

$$\mathbf{f}_{env} = (\mathbf{f}_{wind} + \mathbf{f}_{wave} + \mathbf{f}_{curr})CA_{dvn} \tag{7}$$

where  $f_{wind} = [X_{wind}, Y_{wind}, N_{wind}]^{T}$  are the wind loads,  $f_{wave} = [X_{wave}, Y_{wave}, N_{wave}]^{T}$  are the wave loads and  $f_{curr} = [X_{curr}, Y_{curr}, N_{curr}]^{T}$  are the current loads.  $CA_{dyn}$  is a dynamic allowance coefficient, which is an expedient to include dynamic effects in quasi-static DP predictions. Such coefficient can be derived from time-domain simulation on the same or similar ships or from guidelines provided by Classification Societies. Here, use has been made of a  $CA_{dyn}$ =1.25, as prescribed by DNV-GL [3].

The conventional formulations for environmental loads use non-dimensional coefficients as a function of the calculation heading angles vector  $\chi$ . This approach allows for use the same coefficients on similar ships, without the need to assess with direct methods or model tests the loads. The application is straightforward for wind and current loads, leading to the following formulations:

$$\mathbf{f}_{wind} = \begin{bmatrix} X_{wind} \\ Y_{wind} \\ N_{wind} \end{bmatrix} = \frac{1}{2} \rho_{air} V_w^2 \begin{bmatrix} A_T \\ A_L \\ A_L L_{OA} \end{bmatrix} \mathbf{C}_{wind} (\mathbf{\chi})$$
(8)

$$f_{curr} = \begin{bmatrix} X_{curr} \\ Y_{curr} \\ N_{curr} \end{bmatrix} = \frac{1}{2} \rho_{water} V_c^2 \begin{bmatrix} S \\ S \\ SL_{PP} \end{bmatrix} C_{curr}(\chi)$$
 (9)

where  $C_{wind} = [C_{X_w}, C_{Y_w}, C_{N_w}]^T$  and  $C_{curr} = [C_{X_c}, C_{Y_c}, C_{N_c}]^T$  are the wind and current non-dimensional coefficients, respectively. The other quantities in Eq.(8) and Eq.(9) are:  $\rho_{air}$  the air density,  $\rho_{water}$  the water density,  $V_w$  is the wind speed,  $V_c$  is the current speed,  $A_T$  is the transverse exposed area to wind,  $A_L$  is the lateral exposed area to wind,  $S_v$  is the hull wetted surface,  $S_v$  is the length between perpendiculars and  $S_v$  is the ship's overall length.

The wave forces require a different description of non-dimensional coefficients and calculation process compared to wind and current loads. For quasi-static DP calculations, the wave loads are described by means of mean drift forces, representative of an irregular and usually long-crested sea for specific couples of  $H_s$  and  $T_p$ . Mean drift forces can be measured through dedicated model scale experiments or derived from the quadratic transfer functions (QTF) calculated by diffraction theory. Then, modelling the irregular sea with a spectrum for a specific  $H_s$  and  $T_p$  values, the wave loads result in:

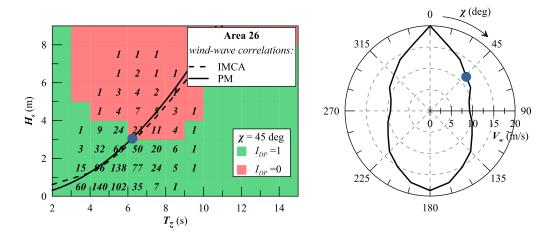
$$\mathbf{f}_{wave} = \begin{bmatrix} X_{wave} \\ Y_{wave} \\ N_{wave} \end{bmatrix} = \rho_{water} g \begin{bmatrix} \nabla^{\frac{1}{3}} \\ \nabla^{\frac{1}{3}} \\ \nabla^{\frac{2}{3}} \end{bmatrix} \int_{0}^{\infty} \mathbf{C}_{wave} (\boldsymbol{\chi}, \boldsymbol{\omega}) S_{\zeta} (\boldsymbol{\omega}) d\boldsymbol{\omega}$$
(10)

where  $C_{wave} = [C_{X_{wav}}, C_{Y_{wav}}, C_{N_{wav}}]^T$  represent the QTF as a function of the heading angle vector  $\chi$  and circular wave frequency vector  $\omega$ .  $\nabla$  is the vessel volume of displacement, g the acceleration of gravity and  $S_{\zeta}$  is the wave amplitude spectrum expressed as a function of circular wave frequency vector  $\omega$ .

From the given description of environmental loads it results that a single resolution of Eq.(5) optimisation problem requires the definition of a tuple of environmental parameters  $e = [V_w, V_c, H_s, T_p, \chi]^T$  to describe the external loads. This is valid assuming that the environmental loads are concurrent within each other, as common practice for quasistatic calculations. Another simplification for quasi-static analysis concerns the current speed  $V_c$ , which is supposed to be constant through all the combination of other environmental parameters. In this study,  $V_c$ =0.75 m/s is used as reference value for current loads.

#### **DP** operability

The conventional quasi-static analysis considers fixed combinations of wind speed and wave parameters through a deterministic relationship. The adoption of such kind of environment definition leads to the determination of the so-called DP system capability through consecutive quasi-static calculations monotonically increasing  $V_w$ . The capability



**FIGURE 3.** Scatter diagram approach representation for one heading  $\chi$  and correlation with a DP capability plot.

represents the maximum sustainable wind speed  $V_{w_{max}}$  the DP system can face, using the on-board actuators, at each heading angle  $\chi$ . The obtained  $V_{w_{max}}$  as a function of  $\chi$  is then reported in a conventional polar plot called Capability Plot. To capture all the possible combinations of wave parameters occurring in sea area, an alternative method is necessary.

A valid methodology is provided by the scatter diagram approach [6]. Instead of performing basic DP capability plot predictions, the scatter diagram approach allows the execution of DP calculations for each combination of  $H_s$  and  $T_p$  of the operational area. Calculations can be carried out for the heading vector  $\chi$ , evaluating whether the DP system is able or not to keep the vessel in position with the resulting sea environment. This strategy allows for a novel evaluation of the DP system performances on a specific sea environment, enhancing the concept of DP capability in favor of the direct quantitative evaluation of the operability of the system. The operability results from the following formulation:

$$OP_{DP} = \sum_{i=1}^{N_{\chi}} f_{\chi_i} \sum_{j=1}^{N_{H_s}} \sum_{k=1}^{N_{T_p}} f_{w_{jk}} I_{DP_{ijk}}$$
(11)

where  $f_{\chi} = [f_{\chi_1}, \cdots, f_{\chi_{N_{\chi}}}]^{\mathrm{T}} \in \mathbb{R}^{N_{\chi}}$  is the occurrence of the  $N_{\chi}$  headings considered in the calculation,  $f_w = [f_{w_{11}}, \cdots, f_{w_{1N_{T_p}}}; \cdots; f_{w_{N_{H_s}}1}, \cdots, f_{w_{N_{H_s}N_{T_p}}}]^{\mathrm{T}} \in \mathbb{R}^{N_{H_s} \times N_{T_p}}$  is the joint occurrence of  $H_s$  and  $T_p$ , provided by a scatter diagram having a granularity of  $N_{H_s}$  wave heights and  $N_{T_p}$  wave periods. The last element to define in Eq.(11) is the matrix  $I_{DP} \in \mathbb{R}^{N_{H_s} \times N_{T_p}}$ , which is containing the results of the single quasi-static analysis on the scatter diagram cells in the following form:

$$I_{DP_{ijk}} = \begin{cases} 1 & \text{if a feasible solution to optimisation problem in Eq.(5) exsist (the DP system holds position)} \\ 0 & \text{otherwise} \end{cases}$$
 (12)

However, a scatter diagram covers only combinations of wave parameters ( $H_s$  and  $T_p$ ), without giving information concerning wind speed  $V_w$ . The discrete nature of Eq.(11) favors the evaluation of a single  $V_w$  for each cell of the scatter diagram, adopting a deterministic and simplified procedure derived from the Pierson-Moskowitz wave spectrum [15]. Such simplification allows for fast evaluation of  $OP_{DP}$ , however does not properly take into account the statistic of  $V_w$  in the geographic sea area. Figure 3 shows an example of results provided with a scatter diagram approach and its correlation with the DP capability plots obtained with Pierson-Moskowitz wind-wave correlation.

#### ENVIRONMENTAL MODELING WITH TRIVARIATE JOINT DISTRIBUTIONS

The process described by the scatter diagram approach can be enhanced by changing the description of the areaspecific long-term environmental conditions. The availability of environmental data derived from local measurements or forecast models allows the definition of joint probability distributions for wind and waves. In the specific, dealing with the necessity to define  $V_w$ ,  $H_s$  and  $T_p$  a trivariate joint distribution should be defined.

A possible formulation for the probability density function of a trivariate joint  $V_w$ - $H_s$ - $T_p$  distribution is as follows:

$$f_{V_w,H_s,T_p}(v_w,h_s,t_p) = f_{V_w}(v_w) f_{H_s|V_w}(h_s,v_w) f_{T_p|V_w,H_s}(t_p,h_s,v_w)$$
(13)

where  $v_w$ ,  $h_s$  and  $t_p$  are three random variables in  $(0, +\infty)$  needed to define the joint distribution. Eq.(13) is composed by a marginal distribution  $f_{V_w}$  for the wind speed, a conditional distribution  $f_{H_x|V_w}$  for the wave height and another conditional distribution  $f_{T_n|V_w,H_s}$  for the wave period. The marginal distribution for  $V_w$  can be modelled with a two parameters Weibull distribution [5]:

$$f_{V_w}(v_w) = \frac{\beta_v}{\eta_v} \left(\frac{v_w}{\eta_v}\right)^{\beta_v - 1} e^{-\left(\frac{v_w}{\eta_v}\right)^{\beta_v}}$$
(14)

where  $\beta_{\nu}$  and  $\eta_{\nu}$  are the shape and scale parameter of the distribution, which are varying depending on the geographic site environmental statistics. A two parameters Weibull model can be employed also for the conditional  $f_{H_8,V_w}$ distribution:

$$f_{H_s|V_w}(h_s, v_w) = \frac{\beta_h}{\eta_h} \left(\frac{h_s}{\eta_h}\right)^{\beta_h - 1} e^{-\left(\frac{h_s}{\eta_h}\right)^{\beta_h}}$$
(15)

This formulation is analogue to Eq.(14), but shape and scale parameters are modelled as power functions of  $v_w$  as follows:

$$\beta_h = a_1 + a_2 v_w^{a_3} \tag{16a}$$

$$\eta_h = b_1 + b_2 v_w^{b_3} \tag{16b}$$

such modelling grants the dependence of Eq.(15) from both  $v_w$  and  $h_s$ .

The conditional distribution  $f_{T_n|H_s,V_w}$  is usually modelled with a log-normal distribution:

$$f_{T_p|H_s,V_w}(t_p, h_s, v_w) = \frac{1}{\sqrt{2\pi}\sigma_{T_s}t_p} e^{-\frac{1}{2}\left(\frac{\ln t_p - \mu_{T_p}}{\sigma_{T_p}}\right)^2}$$
(17)

where the standard deviation  $\sigma_{T_p}$  and the mean value  $\mu_{T_p}$  of  $\ln t_p$  are expressed with the following functions of  $v_w$  and  $h_s$ :

$$\mu_{T_p} = \ln \frac{\mu_{T_p}^*}{\sqrt{\nu_{T_p}^2 + 1}} \tag{18a}$$

$$\sigma_{T_p} = \sqrt{\ln\left(v_{T_p}^2 + 1\right)} \tag{18b}$$

$$V_{T_p} = c_1 + c_2 e^{c_3 h_s} (18c)$$

$$\mu_{T_p}^* = \overline{T_p} \left[ 1 + \theta \left( \frac{v_w - \overline{v_w}}{\overline{v_w}} \right)^{\gamma} \right]$$
 (18d)

$$\overline{T_p} = d_1 + d_2 h_s^{d_3}$$

$$\overline{v_w} = e_1 + e_2 h_s^{e_3}$$
(18e)
(18f)

$$\overline{v_w} = e_1 + e_2 h_s^{e_3} \tag{18f}$$

Such modelling of the trivariate joint distribution with Eq.(14), Eq.(15) and Eq.(17) requires the definition of a set of parameters  $\mathbf{q} = [\eta_{\nu}, \beta_{\nu}, a_1, a_2, a_3, b_1, b_2, b_3, c_1, c_2, c_3, d_1, d_2, d_3, e_1, e_2, e_3, \theta, \gamma]^T$  which are site-specific and determined by fitting procedures on environmental data.

Employing and environmental modelling with Eq.(13) does not allow anymore to have unique combinations of  $V_w$ ,  $H_s$  and  $T_p$ , necessary to apply the scatter diagram approach and evaluate  $OP_{DP}$  according to Eq.(11). It is then necessary to develop a different method for the DP operability evaluation with joint distributions of environmental parameters.

#### DP OPERABILITY AS A MONTE CARLO PROCESS

The adoption of the previously described enhanced environmental modelling with trivariate joint distributions requires a different approach to DP operability calculation. A possible solution is evaluate DP operability as a non-deterministic multidimensional Monte Carlo (MC) integration. An approximated MC integral has the following formulation:

$$\int_{\Omega} f(\boldsymbol{x}) d\boldsymbol{x} \approx \frac{1}{N_s} \int_{\Omega} d\boldsymbol{x} \sum_{i=1}^{N_s} f(\boldsymbol{x}_i)$$
(19)

where  $\Omega \in \mathbb{R}^m$  is an m-dimensional probability space,  $x \in \Omega$  is a matrix of m independent random variables having  $N_s$  elements each. In case  $\Omega$  is an unit hypercube  $(0,1)^m$ , then the term  $\int_{\Omega} dx$  in Eq.(19) is equal to 1. Then, considering uniform random variables  $U \sim \mathbb{U}(0,1)$ , Eq.(19) can be rewritten as:

$$\int_{\Omega} f(\boldsymbol{x}) d\boldsymbol{x} \approx \frac{1}{N_s} \sum_{i=1}^{N_s} f(\boldsymbol{U}_i)$$
 (20)

Using the trivariate joint distribution in Eq.(13) and modelling  $v_w$ ,  $h_s$  and  $t_p$  as uniform random variables U, Eq.(11) can be adapted to an MC integral formulation:

$$OP_{DP} = \sum_{i=1}^{N_{\chi}} f_{\chi_i} \frac{1}{N_s} \sum_{i=1}^{N_s} f_{V_w, H_s, T_p} \left( U_{v_w}, U_{h_s}, U_{t_p} \right) I_{DP_j}$$
(21)

The  $OP_{DP}$  determination according to Eq.(21) maintains a discrete modelling of the heading angles, implying the resolution of a dedicated integration on each  $\chi_i$  angle. Such procedure is useful if the  $OP_{DP}$  for specific headings is of interest. Otherwise, aiming at the determination of the global operability of the DP system, the modelling of the heeding can be also included in the integration process, considering an additional random variable. The MC integral can be than rewritten as:

$$OP_{DP} = \frac{1}{N_s} \sum_{i=1}^{N_s} f_{\chi} \left( \mathbf{U}_{\chi} \right) f_{V_w, H_s, T_p} \left( \mathbf{U}_{v_w}, \mathbf{U}_{h_s}, \mathbf{U}_{t_p} \right) I_{DP_i}$$

$$(22)$$

where  $f_{\chi}$  is a probability density function of headings. The structure of Eq.(22) considers the heading distributions independent from the joint environmental parameters distribution. The evaluation of operability with Eq.(22) requires the evaluation of the tuple of environmental parameters previously introduced; however, being the current speed  $V_c$  fixed in the calculation assumption, the reduced tuple  $e^* = [V_w, H_s, T_p, \chi]^T$  needs to be generated from the pertinent distributions. The procedure for MC integration uses a direct sampling from U with pseudo-random numbers [16], then, dealing with non-uniform distributions, the associated random variables are derived from the inversion of the reference cumulative density function. Considering a generic variable with cumulative F, being U uniform in [0,1], the inverse  $F^{-1}(U)$  is distributed according to F.

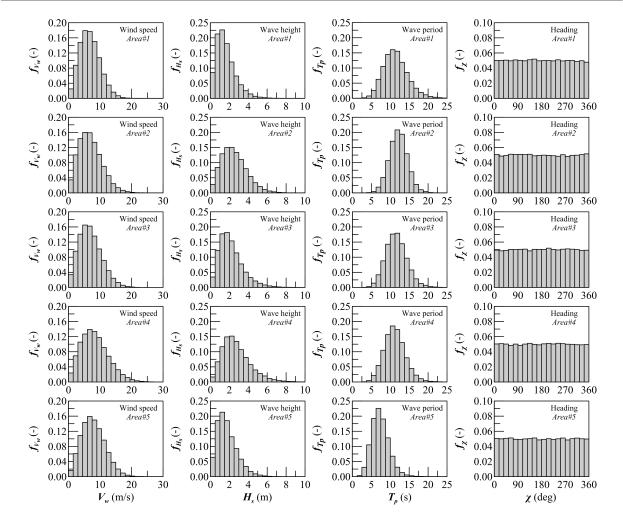
However, the adoption of pseudo-random numbers generates different combinations of parameters in  $e^*$  during each calculation, leading to different final integral values and convergence histories. As the approximated integral converges to an exact value as  $N_s$  increases without upper bounds, then the process is subject to uncertainties. It is then necessary to use a sufficiently large number of samples that ensures the matching of a required confidence level for the solution. This can be achieved by calculating a Confidence Interval (CI) across multiple repetitions  $N_r$ . In the next section the described process is applied to the reference OSV presented in next section for 5 different areas, introducing convergence criteria suitable for offshore operations.

#### MC PROCESS APPLICATION ON THE REFERENCE OSV

The above-described process is here applied on the reference OSV vessel. The first part of the process concerns the environmental modelling. In this example, 5 different geographic sea areas (*Area#1*, *Area#2*, *Area#3*, *Area#4* and *Area#5*) are considered, having the parameters q needed to define the trivariate distribution of Eq.(13) as listed in Table III. The areas are distributed across the European Atlantic coast and the North Sea and are relevant for the installation of offshore wind and waves energy devices [17].

**TABLE III.** Parameters q for the trivariate joint distribution on five reference geographic sea areas.

	$ f_{V_w} \operatorname{Eq.}(14) $ $f_{H_s V_w} \operatorname{Eq.}(15)$					$f_{T_p V_w,H_s}$ Eq.(17)													
Area	10	$eta_{ u}$	$a_1$	$a_2$			$b_2$	$b_3$	$c_1$	$c_2$			$d_2$		$e_1$	$e_2$	$e_3$	$\theta$	γ
1	2.262	7.635	1.894	0.012	1.741	0.929	0.024	1.827	5.0	5.883	0.201	2.0	3.947	0.620	-0.002	0.341	-0.186	-0.268	1.0
2	2.002	7.866	1.643	0.093	1.000	1.963	0.031	1.644	5.0	5.970	0.223	1.0	4.055	0.466	0.030	0.234	-0.221	-0.143	1.0
3	2.050	7.859	2.044	0.034	1.375	1.323	0.032	1.757	8.0	2.600	0.409	1.8	3.478	0.667	0.002	0.298	-0.166	-0.233	1.0
4	2.029	9.409	2.136	0.013	1.709	1.816	0.024	1.787	8.0	1.938	0.486	2.5	3.001	0.745	-0.001	0.316	-0.145	-0.255	1.0
5	2.299	8.920	1.755	0.184	1.000	0.534	0.070	1.435	5.563	0.798	1.0	3.5	3.592	0.735	0.050	0.388	-0.321	-0.477	1.0

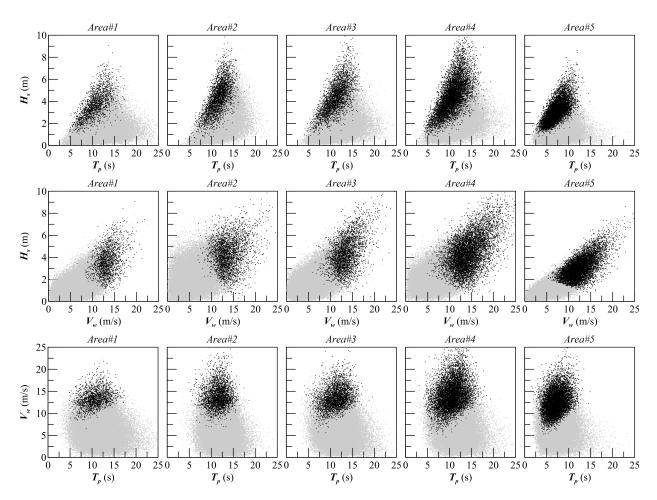


**FIGURE 4.** Marginal distributions of the environmental variables and heading angles generated for the 5 reference geographic sea areas.

However, the application of Eq.(22) to evaluate  $OP_{DP}$  for the reference OSV requires, besides the environmental conditions, to define the function  $f_{\chi}$  describing the distribution of headings  $\chi$ , and then determining  $e^*$  for the DP analyses. Here, the process is applied considering  $f_{\chi}$  as a uniform random variable  $U_{\chi} \sim \mathbb{U}(0,1)$  independent from the three random variables  $v_w$ ,  $h_s$  and  $t_p$  used for the trivariate joint distribution of environmental parameters. That means all the angles  $\chi$  in [0,360] are considered equiprobable and independent from other simulation parameters.

Considering a sample size  $N_s = 10^5$ , the determination of  $e^*$  is performed according to the following steps:

- 1. Generation of  $\Omega = [U_{\chi}, U_{V_w}, U_{H_s}, U_{T_p}]^{\mathrm{T}}$  with pseudo-random number sequences with sample size  $N_s$ .
- 2. Direct determination of  $\chi$  from  $U_{\chi}$ .



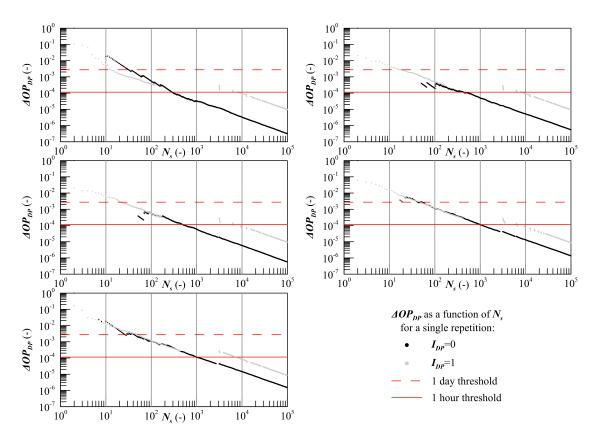
**FIGURE 5.**  $I_{DP}$  values as a function of  $H_s - T_p$  (top),  $H_s - V_w$  (middle) and  $V_w - T_p$  (bottom) for the five reference areas.

- 3. Inversion of  $f_{V_w}$  (Eq.(14)) to obtain the wind speed random variable  $v_w$ .
- 4. Inversion of  $f_{H_s|V_w}$  (Eq.(15)) using  $v_w$  to obtain the significant wave height random variable  $h_s$ .
- 5. Inversion of  $f_{T_p|V_w,H_s}$  (Eq.(17)) using  $v_w$  and  $h_s$  to obtain the peak period random variable  $t_p$ .
- 6. Determination of  $e^*$  for all the  $N_s$  samples.

The results of the sampling process on the reference ship for the five geographic sea areas is presented in Figure 4. Here the marginal distributions of  $v_w$ ,  $h_s$ ,  $t_p$  and  $\chi$  are reported in form of histograms. The resulting marginal distributions highlight the differences between the environmental conditions in the five sea areas. Figure 4 shows also the uniform sampling of headings  $\chi$ , highlighting the differences in the final uniform distributions provided by the 5 independent random generation processes used to generate the environmental conditions.

The resulting set of  $e^*$  inputs is then used to perform  $N_s$  DP analyses solving the nonlinear optimisation problem in Eq.(5). Figure 5 shows the results of the DP analyses, visualising the  $I_{DP}$  function for all the  $N_s$  calculation per each geographic sea area. The results are reported considering the pairwise comparison of  $V_w$ ,  $H_s$  and  $T_p$ . According to Eq.(12), the function  $I_{DP}$  has only value 0 or 1 either the DP system holds or not the position. Therefore, in Figure 5 it is directly possible to identify the critical combination of parameters for the DP system in the five different areas. The representation of  $I_{DP}$  as a function of  $T_p$  and  $H_s$  gives the same insight of the scatter diagram approach but in this case, being the wind no more fixed for the  $H_s$ - $T_p$  couples and being the results comprehensive of  $\chi$  variations there is no clear separations between regions with  $I_{DP} = 1$  and  $I_{DP} = 0$ .

The evaluation of  $OP_{DP}$  requires to check whether the MC integration process has reached a sufficient level of convergence. This can be performed monitoring the relative differences between  $OP_{DP}$  values at consecutive integration



**FIGURE 6.** Convergence of MC process on a single  $OP_{DP}$  calculation for the five reference sea areas.

steps:

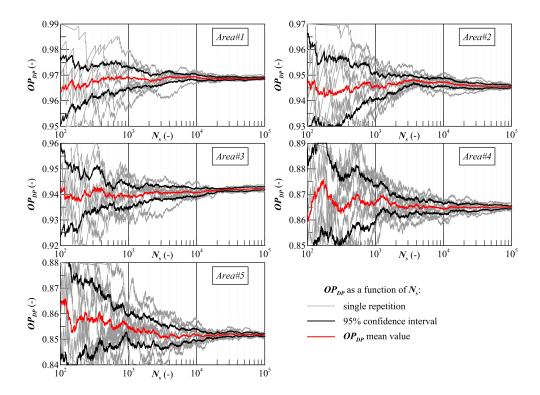
$$\Delta OP_{DP_i} = \left| \sum_{j=1}^{i} \frac{I_{DP_j}}{i} - \sum_{j=1}^{i-1} \frac{I_{DP_j}}{i-1} \right|$$
 (23)

It is than necessary to compare the  $\Delta OP_{DP}$  with a reference threshold value significant for the analysed problem convergence. Being the  $OP_{DP}$  a quantity defined between 0 and 1 that indicates the fraction of year a vessel could operate in a sea area at a given encounter angle, the convergence should be related to the time unit used to quantify the operability. In case it is quantified in days, considering 1 day as convergence threshold, the convergence can be reached when  $\Delta OP_{DP}$  approaches  $2.74\cdot10^{-3}$ . Considering a threshold of 1 hour, then the  $\Delta OP_{DP}$  of reference is  $1.144\cdot10^{-4}$ .

Figure 6 shows the  $OP_{DP}$  variations with  $N_s$  for the 5 reference geographic sea areas in logarithmic scale. It is possible to observe in all the subplots two distinct series of data, one corresponding to the results with  $I_{DP} = 1$  and the other with  $I_{DP} = 0$ . As  $I_{DP}$ , the integrating function, has only two possible discrete values, the  $\Delta OP_{DP}$  curve has two distinct trends and, consequently, the convergence should be checked on the higher sequence of points in the reference  $N_s$  range. Figure 6 shows also the two convergence thresholds corresponding to 1 day and 1 hour. It can be observed that for the 5 areas the 1 day threshold is reached with less than 100 samples, while satisfying the 1 hour threshold requires almost  $10^4$  samples for all the tested cases.

However, to consider the convergence of a single calculation may be misleading for the MC integration due to the fact that the MC process is subject to uncertainties. Therefore, for the estimation of the  $OP_{DP}$  it is advisable to perform multiple repetitions  $N_r$  to evaluate the DP operability within a confidence interval CI. Generally, when  $N_r$  is sufficiently high, a CI can be described by a normal distribution. Here, the  $OP_{DP}$  is analysed with  $N_r = 10$ , suggesting to employ the following formulation for CI:

$$CI(c) = \mu \pm t \frac{\sigma}{\sqrt{N_r}}$$
 (24)



**FIGURE 7.** Solution history of  $OP_{DP}$  in the five selected areas with  $N_r = 10$ .

**TABLE IV.**  $OP_{DP}$  mean values  $\mu$  and CI for the five reference sea area considering  $N_r = 10$  and different  $N_s$  values.

$N_s$		Area#1	Area#2	Area#3	Area#4	Area#5
$1 \cdot 10^3$	μ	0.96910	0.94550	0.93930	0.86680	0.85470
	CI(95%)	$\pm 3.76E-3$	$\pm 4.75E-3$	$\pm 4.32E-3$	$\pm 6.40E-3$	$\pm 4.45E-3$
5 103	$\mu$	0.96906	0.94726	0.94086	0.86556	0.85094
$5 \cdot 10^3$	CI(95%)	$\pm 1.55E-3$	$\pm 1.66E-3$	$\pm 2.14E-3$	$\pm 2.33E-3$	$\pm 3.19E-3$
$1 \cdot 10^4$	$\mu$	0.96909	0.94719	0.94075	0.86424	0.85161
	CI(95%)	$\pm 1.11E-3$	$\pm 1.53E-3$	$\pm 1.58E-3$	$\pm 1.94E-3$	$\pm 2.24E-3$
$5 \cdot 10^4$	$\mu$	0.96844	0.94572	0.94198	0.86466	0.85197
	CI(95%)	$\pm 5.07E-4$	$\pm 6.62E-4$	$\pm 8.36E-4$	$\pm 9.96E-4$	$\pm 6.94E-4$
$1 \cdot 10^5$	$\mu$	0.96844	0.94550	0.94244	0.86488	0.85162
	CI(95%)	$\pm 3.92E-4$	$\pm 4.99E-4$	$\pm 5.83E-4$	$\pm 5.14E-4$	$\pm 5.28E-4$

where  $\mu$  is the mean value of  $OP_{DP}$  in the  $N_r$  repetition,  $\sigma$  is the repetitions variance and t is the inverse cumulative density function of the Student t-distribution with confidence interval c and  $N_r = 1$  degrees of freedom. Here, a 95% CI has been selected for the  $OP_{DP}$  estimation.

Figure 7 shows the  $OP_{DP}$  value as a function of  $N_s$  on the 5 reference geographic sea areas. The drawing reports the 10 repetitions together with the variation of the mean value  $\mu$ , and the confidence interval CI. It can be observed that, as expected, the confidence interval CI reduces by increasing  $N_s$  for all the reference cases. However, comparing the CI values with the 1 day and the 1 hour thresholds, the convergence of the process is different than what observed for a single repetition. In fact, none of the selected cases reaches a CI beyond the 1 hour threshold, even considering all  $N_s = 10^5$  samples. The 1 day threshold is reached for different  $N_s$  values depending on the sea area of reference, ranging from 1540 in Area#I to 6327 in Area#5. Table IV reports the mean value  $\mu$  and the CI for the  $OP_{DP}$  evaluated in the 5 areas with different  $N_s$ . From the results it can be derived that the areas where the CI goes beyond the 1 day thresholds with less samples corresponds to the ones having higher  $OP_{DP}$ .

The  $OP_{DP}$  values obtained in the considered geographic sea areas are different between each other of more than 10%, highlighting the importance to perform DP predictions specific for the operational area of interest for the vessel.

This area-specific DP prediction increases the calculation time compared to a conventional capability analysis, as any single repetition for one sea area requires about 20 minutes of calculation time with  $N_s = 10^5$ . However, such a calculation effort is acceptable for DP analyses and ship design purposes. An improvement of the procedure can be obtained using different sample techniques for the joint distributions [18] that will speed up the convergence of the integral, thus decreasing the calculation time.

#### CONCLUSION

The present work implements a methodology based on MC integration to perform area specific DP predictions, starting from the modelling of environmental parameters with joint trivariate distributions. The adoption of such trivariate distributions does no more allow for the execution of a standard DP capability analysis, but requires the development of a specific process to determine the operability of the DP system in the geographic sea area. The application of the process on a reference OSV for five different sea areas highlight the importance of perform dedicated prediction for area-specific operations, as the operability value varies of more than 10% for the tested cases. The analyses performed on the convergence of the solution highlights that with the adopted number of samples and repetitions, operability can be determined within a confidence interval of 1 day in one year of operations. The area-specific DP prediction can be further improved by performing additional studies on the sampling process employed to determine the environmental random variables, aiming at a reduction of the results variance, thus of the calculation time. Finally, the newly proposed MC method increases the reliability of the environmental modelling for DP predictions, which is a significant improvement for practical engineering application of offshore industry.

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