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Analyzing the dynamic response of a rotor system under uncertain parameters by Polynomial Chaos Expansion

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Abstract

In this paper, the quantification of uncertainty effects on response variability in rotor systems is investigated. To avoid the use of Monte Carlo simulation (MCS), one of the most straightforward but computationally expensive tools, an alternative procedure is proposed. Monte Carlo Simulation builds statistics from responses obtained from sampling uncertain inputs by using a large number of runs. However, the method proposed here is based on the stochastic finite element method (SFEM) using polynomial chaos expansion (PCE).

The efficiency and robustness of the method proposed is demonstrated through different numerical simulations in order to analyze the random response against uncertain parameters and random excitation to assess its accuracy and calculation time.

Keywords: dynamics, rotor, uncertainties

1 Introduction

In rotordynamics, two types of uncertainty on dynamic systems are of particular interest. The first of these can derive from variations in mechanical properties (such as mass, stiffness and geometrical imperfections) due to manufacturing errors [Lalanne and Ferraris (1990); Friswell and Mottershead (1995); Erich (1992); Childs (1993); Yamamoto and Ishida (2001)]. Besides this type of structural uncertainties, external and internal forcing functions can also be random.

Numerous methods have been used to quantify physical uncertainties in a variety of computational problems like the perturbation method, the Monte Carlo Simulations and the Polynomial Chaos Expansion [Ghanem and Spanos (1991)]. The perturbation method based on the expansion of random quantities into Taylor series [Nayfeh (1973)] and the Neumann method based on Neumann series expansion [Benaroya and Rehak (1988); Yamazaki et al. (1988)] provide acceptable results for small random fluctuations, then they are not adapted here for solving a dynamic problem in the case of an excitation frequency close to a resonance frequency. The direct Monte Carlo Simulations, which is the most straightforward and frequently used approach, is adapted to include uncertainties in a deterministic finite element model, by generating n independent samples of the random parameter. Then it leads to solve the deterministic problem n times in order to obtain n samples of the response vector and so the statistics of the response. Due to the slow convergence rate of this method, a very high number of samples is necessary then if solving the deterministic problem is already computationally intensive, the computational costs of the method can become impractical. Particularly, rotordynamics problems are quite complex to solve in a deterministic sense. The polynomial chaos expansion associated with a Galerkin projection so-called stochastic finite element method [Ghanem and Spanos (1991)] has shown to be a successful approach to solve uncertainty quantification problems. It represents the stochastic processes and variables in a set of orthogonal bases of random variables. The polynomial chaoses are from the homogeneous chaos theory of Wiener [Wiener (1938)] and the original polynomial chaos expansion [Ghanem and Spanos (1991)] used a mean-square convergent expansion as multidimensional Hermite polynomials of normalized Gaussian variables. Since the Hermite polynomials are orthogonal with respect to the Gaussian measure, the homogeneous polynomial chaos can achieve optimal exponential convergence for Gaussian inputs [Ghanem (1999)]. This last method then seems the most adapted to study the influence of the uncertainties on the parameters of the rotor structure on the response.

The present paper is organized as follows: firstly, we present the rotor system after which a brief explanation is given of the Stochastic Finite Element Method [Ghanem and Spanos (1991)] for the solution of mechanical problems with several random characteristics. Secondly, expansions of the operator of random material properties and of the random external forcing function on the chaos basis are explained and studied for application to the rotordynamics problem. The Polynomial Chaos Expansion procedure is illustrated by different numerical examples that include the most common sources of randomness in a rotordynamics problem (such as physical and geometric parameters). Thirdly, the results obtained by applying the Polynomial Chaos Expansion (PCE) procedure are compared with those evaluated by Monte Carlo Simulation (MCS) whose costs become prohibitive for large finite element models with large numbers of design parameters. Finally, the efficiency and robustness of the method proposed is demonstrated through several numerical simulations of the effects of uncertainties and orders of polynomial chaos.

2 Rotor Model

The system under study is illustrated in Figure 1. The rotor consists of a rotor shaft with two discs. The shaft is discretized into 10 Timoshenko beam finite elements with four degrees of freedom at each node [Lalanne and Ferraris (1990); Friswell and Mottershead (1995)] and a constant circular section. All the values of the parameters are given in Table 1. The beam element model is given by

$$
[\mathbf{M}^e]{\mathbf{x}^e} + ([\mathbf{C}^e] + \omega[\mathbf{G}^e]){\mathbf{x}^e} + [\mathbf{K}^e]{\mathbf{x}^e} = {\mathbf{q}^e}
$$
 (1)

where the vector $\{q^e\}$ defines forces applied on the shaft and ω is the rotational speed of the shaft. [M^e] and [G^e] are the mass and gyroscopic matrices of the shaft element. [K^e] and [C^e] are the elementary stiffness and damping matrices. These matrices are described in A.

The model of the rigid discs is given by

$$
[\mathbf{M}^d]\{\mathbf{x}^d\} + \omega[\mathbf{G}^d]\{\mathbf{x}^d\} = \{\mathbf{q}^d\}
$$
 (2)

where $[\mathbf{M}^d]$ and $[\mathbf{G}^d]$ are the mass and gyroscopic matrices of the disc. These matrices will be described in the following part of the paper. The vector $\{q^d\}$ defines the unbalance forces due to an eccentric mass. For the degree-of-freedom $[v \quad w \quad \theta \quad \psi]^T$ (see Figure 1), the unbalance forces are given by

$$
\mathbf{q}^d = [m_e d_e \omega^2 \cos(\omega t + \phi) \quad m_e d_e \omega^2 \sin(\omega t + \phi) \quad 0 \quad 0]^T
$$
 (3)

where m_e and d_e are the mass unbalance and the eccentricity respectively. ϕ and ω define the initial angular position in relation to the z-axis and the rotational speed of the rotor. Finally, discrete stiffness components are located at either end of the shaft. After assembling the shaft elements and the rigid discs, the equation of motion for the complete rotor system is defined as follows:

$$
[\mathbf{M}]\{\ddot{\mathbf{x}}\} + ([\mathbf{C}] + \omega[\mathbf{G}])\{\dot{\mathbf{x}}\} + ([\mathbf{K}] + [\mathbf{K}_b])\{\mathbf{x}\} = \{\mathbf{q}\}\
$$
(4)

where [**M**] and [**G**] are the mass and gyroscopic matrices of the shaft and the two discs. [**K**] and [**C**] are the stiffness and damping matrices of the shaft. $[\mathbf{K}_b]$ is the stiffness matrix of the bearings. $\{\mathbf{q}\}\$ is the unbalance forces of the complete rotor system. Considering that the unbalance force can be written as $\{q\} = \{Q\}e^{i\omega t}$, the response vector may be assumed to be $\{x\} = \{X\}e^{i\omega t}$. By using Equation (4), the system governing the equation in the frequency domain is given by

$$
[\mathbf{A}(\omega)]\{\mathbf{X}(\omega)\} = \{\mathbf{Q}(\omega)\}\tag{5}
$$

where

$$
[\mathbf{A}(\omega)] = -\omega^2[\mathbf{M}] + i\omega([\mathbf{C}] + \omega[\mathbf{G}]) + [\mathbf{K}] + [\mathbf{K}_b]
$$
\n(6)

In the following part of the paper, the frequency dependence will be omitted to simplify the notation and $[\mathbf{A}(\omega)]$ will be noted as $[\mathbf{A}]$.

3 Stochastic model

In rotordynamics, uncertainties on dynamic responses can occur due to manufacturing inaccuracies related to mechanical properties such as mass, stiffness, damping and geometrical imperfections. Moreover, forcing functions (external and internal) also lead to considerable uncertainties, so that they have to be taken as random quantities. Therefore, considering in this part $[A], \{X\}$ and $\{Q\}$ as random processes, where argument τ denotes the random character, Equation (5) can then be rewritten in a random way such that

$$
[\mathbf{A}(\tau)]\{\mathbf{X}(\tau)\} = \{\mathbf{Q}(\tau)\}\tag{7}
$$

Figure 1: Rotor system and Shaft finite element

Material and geometrical properties are randomly modeled by $[M(\tau)]$, $[G(\tau)]$, $[K(\tau)]$, $[C(\tau)]$ and $[K_b(\tau)]$, thus Equation (6) becomes

$$
[\mathbf{A}(\tau)] = -\omega^2[\mathbf{M}(\tau)] + i\omega([\mathbf{C}(\tau)] + \omega[\mathbf{G}(\tau)]) + [\mathbf{K}(\tau)] + [\mathbf{K}_b(\tau)]
$$
\n(8)

3.1 The system equation on the Polynomial Chaos basis

The Polynomial Chaos basis is a set of orthogonal bases of random variables, represented in a meansquare convergent expansion by multidimensional Hermite polynomials of normalized Gaussian variables. A rapid overview on the construction of the basis is given below and more details can be found in [Ghanem and Spanos (1991)]. In the following of the paper, the random behavior of each physical or geometrical quantity **M** (scalar or matrix) considered would be sufficiently modeled by using the Karhunen-Loeve expansion implemented in the Galerkin formulation of the finite element method [Ghanem and Spanos (1991)], then we can expand **M** such as

$$
\mathbf{M} = \overline{\mathbf{M}} + \sum_{l=1}^{L} \xi_l \mathbf{M}_l
$$
 (9)

where $\{\xi_1, \dots, \xi_L\}$ is a set of orthonormal random variables, **M** is the mean of **M** and **M**_l is its lth expansion term. The response can be expanded on the Polynomial Chaos basis such that

$$
\left\{ \mathbf{X}(\tau) \right\} = \sum_{j=0}^{\infty} \left\{ \mathbf{X} \right\}_j \Psi_j(\underline{\xi}(\tau)) \tag{10}
$$

Parameters	dimension
Length of shaft	$1~\mathrm{m}$
Diameter of shaft	$0.04 \;{\rm m}$
Position of disc 1	0.6 m
Position of disc 2	0.8 _m
Outer diameter of disc 1	0.2 m
Outer diameter of disc 2	0.4 m
Inner diameter of discs 1 and 2	$0.04 \; \mathrm{m}$
Thickness of discs 1 and 2	0.02 m
Young modulus of elasticity E	2.1 10^{11} Nm ²
Shear modulus G	8.0 10^{10} Nm ²
Poisson ratio ν	0.3
Density ρ	7800 kg m^{-3}
Mass unbalance	0.05 g
Phase unbalance	0°
Eccentricity of the mass unbalance	0.02 m
Damping factor η	0.03

Table 1: Model parameters

where $\Psi_j(\xi(\tau))$ refers to a rearrangement of the p-order finite dimensional orthogonal polynomials in relation to the Gaussian function that forms a complete basis in the space of second-order random variables; $\{X\}_j$ is the unknown deterministic *j*th vector associated with $\Psi_j(\xi(\tau))$ and $\xi = \{\xi_r\}$ [Ghanem and Spanos (1991)]. Finally, the system to be solved, when expanded on the polynomial chaos basis, is

$$
\sum_{j=0}^{\infty} [\mathbf{A}(\tau)] {\mathbf{X}}_{j} \Psi_{j}(\underline{\xi}(\tau)) = {\mathbf{Q}(\tau)} \tag{11}
$$

with random quantities $[A(\tau)]$ and $[Q(\tau)]$ defined by

$$
[\mathbf{A}(\tau)] = \sum_{i=0}^{\infty} [\mathbf{A}]_i \Psi_i(\underline{\xi}(\tau))
$$
\n(12)

with

$$
[\mathbf{A}]_i = -\omega^2 [\hat{\mathbf{M}}]_i + i\omega([\hat{\mathbf{C}}]_i + \omega[\hat{\mathbf{G}}]_i) + [\hat{\mathbf{K}}]_i + [\hat{\mathbf{K}}_b]_i \qquad i = 0, 1, ..., \infty
$$
\n(13)

and

$$
[\mathbf{Q}(\tau)] = \sum_{k=0}^{\infty} {\{\hat{\mathbf{Q}}\}_k \Psi_k(\underline{\xi}(\tau))}
$$
\n(14)

where quantity \dot{Q} denotes the rearrangement of quantity Q in the polynomial chaos basis. Details of the rearrangement will be given in the following section of the paper. It should be noted that each random expansion will be described in Subsection 3.2. Finally, the system to be solved to the subspace spanned by $\{\Psi_k\}_{k=0}^{\infty}$, is

$$
\left(\sum_{i=0}^{\infty} [\mathbf{A}]_i \Psi_i(\underline{\xi}(\tau))\right) \left(\sum_{j=0}^{\infty} {\{\mathbf{X}\}_j \Psi_j(\underline{\xi}(\tau))}\right) = \sum_{k=0}^{\infty} {\{\hat{\mathbf{Q}}\}_k \Psi_k(\underline{\xi}(\tau))}
$$
(15)

which, after doing a Galerkin projection on the polynomial chaos basis, can also be rewritten as

$$
\sum_{i=0}^{\infty} \sum_{j=0}^{\infty} E\{\Psi_i \Psi_j \Psi_k\} [\mathbf{A}]_i {\mathbf{X}}_j = {\mathbf{Q}}_k E\{\Psi_k^2\} \qquad k = 0, 1, ..., \infty
$$
\n(16)

where $E\{\}$ denotes the operation of mathematical expectation. It should be noted that coefficients $E{\Psi_i\Psi_j\Psi_k}$ and $E{\Psi_k^2}$ only have to be calculated once. In practice, the expansion can be truncated after the Pth term where P is the total number of polynomial chaoses used in the expansion excluding the 0th order term and can be determined by

$$
P = 1 + \sum_{s=1}^{p} \frac{1}{s!} \prod_{r=0}^{s-1} (L+r)
$$
\n(17)

and in which p is the order of homogeneous chaos used.

3.2 The random quantities in the stochastic rotor

There are different sources of randomness in the rotordynamics problem studied due to geometric and material parameters. This paper deals with uncertainties modeled by Gaussian random variables that represent the random character of the parameters, such as the Young modulus of the shaft, bearing stiffness, disc diameter and density and the amplitude of the unbalance force. All these random quantities are modeled by using expansion defined in Equation (9) and, for physically strictly positive parameters, the random variables of negative values have been removed.

Stiffness of the shaft From the random character of the Young modulus modeled by one truncated Gaussian random variable ξ_1 , thus, from expansion described in Equation (9), we obtain the relation

$$
E(\tau) = \overline{E}(1 + \delta_E \xi_1) \tag{18}
$$

where \overline{E} and δ_E are respectively the mean value and the variation coefficient of the Young modulus. The detailed deterministic expression of the elementary stiffness matrix of the shaft $[K^e]$ as a function of the Young modulus is described in A. By introducing the random Young modulus defined above and after assembling the elementary stiffness matrices of the shaft, it is easy to find the random expansion of [**K**]

$$
\left[\mathbf{K}\right] = \left[\mathbf{K}\right]_0 + \xi_1[\mathbf{K}]_1\tag{19}
$$

(20)

j		Ψ_j	$E\{\Psi_i^2\}$
	$\frac{p}{0}$	$\mathbf 1$	1
	$\mathbf 1$		$\mathbf 1$
			$\mathbf{1}$
			$\mathbf 1$
			$\mathbf{1}$
			$\mathbf{1}$
123456 789		$\frac{1}{\xi_1}$ ξ_2 ξ_3 ξ_4 ξ_5 ξ_6	$\mathbf{1}$
		ξ_1^2 $\overline{1}$	$\overline{2}$
			$\mathbf{1}$
			$\mathbf{1}$
10			$\mathbf 1$
11		$\xi_1 \xi_2$ $\xi_1 \xi_3$ $\xi_1 \xi_4$ $\xi_1 \xi_5$	$\mathbf{1}$
			$\mathbf{1}$
			\overline{c}
12 13 14		$\begin{array}{r} \xi_1\xi_6\\ \xi_2^2-1\\ \xi_2\xi_3\\ \xi_2\xi_4\\ \xi_2\xi_5\\ \xi_2\xi_6\\ \xi_3^2-1\\ \end{array}$	$\overline{1}$
15			$\mathbf 1$
16			$\mathbf{1}$
17			$\mathbf{1}$
18			\overline{c}
19			$\mathbf{1}$
$\overline{20}$			$\mathbf{1}$
$\overline{21}$			$\mathbf 1$
			\overline{c}
			$\mathbf{1}$
22 23 24 25			$\mathbf{1}$
		$\xi_3\xi_4$ $\xi_3\xi_5$ $\xi_3\xi_6$ $\xi_4\xi_5$ $\xi_4\xi_6$ ξ_5 ξ_6 ξ_6 ξ_6 ξ_6 ξ_6	\overline{c}
26 27			$\mathbf{1}$
			\overline{c}

Table 2: Six-Dimensionnal Polynomial Chaoses and their variance

Table 3: Coefficient $E\{\Psi_i\Psi_j\Psi_k\}$, $E\{\Psi_i\Psi_j\Psi_k\} = E\{\Psi_j\Psi_i\Psi_k\} = E\{\Psi_i\Psi_k\Psi_j\}$, Six-Dimensional Polynomial Chaoses, Chaos order 2

Similarly, the hysteretic damping is defined by

$$
\left[\mathbf{C}^e\right] = \frac{\eta}{\omega} \left[\mathbf{K}^e\right] \tag{21}
$$

where η is the hysteretical damping factor. Assembling elementary damping matrices of the shaft yields:

$$
[\mathbf{C}] = [\mathbf{C}]_0 + \xi_1[\mathbf{C}]_1 \tag{22}
$$

Bearing stiffness For the shaft corresponding to the degree-of-freedom $[v w \theta \psi]^T$, the deterministic elementary stiffness matrix of the bearing is defined as

$$
\begin{bmatrix} \mathbf{K}_b^e \end{bmatrix} = \begin{bmatrix} k_{1x} & 0 & 0 & 0 \\ & k_{1y} & 0 & 0 \\ & & 0 & 0 \\ & & sym & 0 \end{bmatrix}
$$
 (23)

where k_{1x} and k_{1y} are the stiffnesses of the bearing in directions x and y. In this case, it has been chosen to only investigate the randomness of the stiffness on the first bearing in direction x modeled by the truncated Gaussian random variable ξ_2 , which yields:

$$
k_{1x}(\tau) = k_{1x}(1 + \delta_{k_{1x}}\xi_2)
$$
\n(24)

in which k_{1x} and $\delta_{k_{1x}}$ are the mean value and the variation coefficient of the stiffness. Finally, the expression of the assembled stiffness matrix $[K_b]$ is written as

$$
\left[\mathbf{K}_b\right] = \left[\mathbf{K}_b\right]_0 + \xi_2 \left[\mathbf{K}_b\right]_1 \tag{25}
$$

Disc parameters The parameters of the discs should be random. Here, we consider the randomness on the outer diameter $D(\tau)$ and the density $\rho(\tau)$ of one of the discs, which are the geometric and material parameters of the model. Describing them by using two truncated Gaussian random variables ξ_3 and ξ_4 yields

$$
D(\tau) = \overline{D}(1 + \delta_D \xi_3) \tag{26}
$$

$$
\rho(\tau) = \overline{\rho}(1 + \delta_{\rho}\xi_4) \tag{27}
$$

where \overline{D} and $\overline{\rho}$ are the mean values, δ_D and δ_ρ are the variation coefficients of the diameter and the density of the disc respectively. These quantities appear in the definition of the mass and gyroscopic elementary matrices $[\mathbf{M}^d]$ and $[\mathbf{G}^d]$ that are expressed for the disk relative to the degree of freedom $[v \ w \ \theta \ \psi]^{T}$ such that

$$
\begin{bmatrix} \mathbf{M}^d \end{bmatrix} = \begin{bmatrix} m_d(\tau) & 0 & 0 & 0 \\ 0 & m_d(\tau) & 0 & 0 \\ 0 & 0 & I_d(\tau) & 0 \\ 0 & 0 & 0 & I_d(\tau) \end{bmatrix} , \quad \begin{bmatrix} \mathbf{G}^d \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & -I_p(\tau) \\ 0 & 0 & I_p(\tau) & 0 \end{bmatrix}
$$
 (28)

with

$$
m_d(\tau) = \frac{1}{4}\rho(\tau)\pi h(D(\tau)^2 - d^2)
$$
\n(29)

$$
I_d(\tau) = \frac{1}{64} \rho(\tau) \pi h (D(\tau)^4 - d^4) + \frac{1}{48} \rho(\tau) \pi h^3 (D(\tau)^2 - d^2)
$$
 (30)

$$
I_p(\tau) = \frac{1}{32} \rho(\tau) \pi h (D(\tau)^4 - d^4)
$$
\n(31)

in which h and d are the thickness and the inner diameter of the disc respectively. In addition, m_d is the mass of the disk. I_d and I_p are the diametral moment of inertia about any axis perpendicular to the rotor axis and the polar moment of inertia about the rotor axis.

Substiting Equations (26-27) in Equations (29) to (31) leads to the expression of the components of the mass and gyroscopic elementary matrices such that

$$
m_d(\tau) = \sum_{j=0}^4 \sum_{i=0}^1 m_{dij} \xi_3^j \xi_4^i \qquad I_d(\tau) = \sum_{j=0}^4 \sum_{i=0}^1 I_{dij} \xi_3^j \xi_4^i \qquad I_p(\tau) = \sum_{j=0}^4 \sum_{i=0}^1 I_{p_{ij}} \xi_3^j \xi_4^i \qquad (32)
$$

where m_{dij} , I_{dij} and I_{pij} are given in B. The Polynomial Chaos Expansion for $m_d(\tau)$, $I_d(\tau)$ and $I_p(\tau)$, constructed for two random variables ξ_3 and ξ_4 is given by

$$
m_d(\tau) = \sum_{j=0}^{N} m_{dj} \Psi_j(\xi_3, \xi_4) \qquad I_d(\tau) = \sum_{j=0}^{N} I_{dj} \Psi_j(\xi_3, \xi_4) \qquad I_p(\tau) = \sum_{j=0}^{N} I_{pj} \Psi_j(\xi_3, \xi_4) \qquad (33)
$$

in which m_{dij} , I_{dij} and I_{pij} are determined after identification between Equations (32) and (33) using Tables 4 and 5. The number of polynomial chaoses N is deduced from Equation (17) by two random variables: $L = 2$. It should be noted that this identification yields a minimum order $p = 5$. Then, expressing mass and gyroscopic elementary matrices of a disc $[\mathbf{M}^{de}]$ and $[\mathbf{G}^{de}]$ in the polynomial chaos basis yields

$$
[\mathbf{M}^d] = \sum_{j=0}^P [\mathbf{M}^d]_j \Psi_j(\xi_3, \xi_4) \qquad [\mathbf{G}^d] = \sum_{j=0}^P [\mathbf{G}^d]_j \Psi_j(\xi_3, \xi_4)
$$
(34)

where

$$
[\mathbf{M}^{d}]_{j} = \begin{bmatrix} m_{dj} & 0 & 0 & 0 \\ 0 & m_{dj} & 0 & 0 \\ 0 & 0 & I_{dj} & 0 \\ 0 & 0 & 0 & I_{dj} \end{bmatrix} , \quad [\mathbf{G}^{d}]_{j} = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & -I_{pj} \\ 0 & 0 & I_{pj} & 0 \end{bmatrix}
$$
(35)

Finally, the assembled mass and gyroscopic matrices are given by

$$
[\mathbf{M}] = \sum_{j=0}^{P} [\mathbf{M}]_j \Psi_j(\xi_3, \xi_4) \quad \text{and} \quad [\mathbf{G}] = \sum_{j=0}^{P} [\mathbf{G}]_j \Psi_j(\xi_3, \xi_4)
$$
 (36)

$\dot{\jmath}$	\overline{p}	Ψ_j	$E\{\Psi_i^2\}$
$\boldsymbol{0}$	0	$\mathbf{1}$	
$\overline{1}$	1	ξ_3	$\mathbf{1}$
\overline{c}		ξ_4	$\mathbf{1}$
$\frac{3}{4}$	$\overline{2}$	$\sqrt{\xi_3^2-1}$	$\overline{2}$
		$\xi_3\xi_4$	$\mathbf{1}$
		$\xi_4^2 - 1$	$\overline{2}$
$\overline{6}$	3	$\frac{\xi_3^3 - 3\xi_3}{\xi_3^2 \xi_4 - \xi_4}$	6
$\overline{7}$			$\overline{2}$
8		$\xi_3^{\dot{1}}\xi_4^2 - \xi_3$	$\overline{2}$
$\overline{9}$		$\xi_4^3 - 3\xi_4$	6
10	4	$\xi_3^4 - 6\xi_3^2 + 3$ $\xi_3^3\xi_4 - 3\xi_3\xi_4$	$\overline{24}$
11			$\overline{6}$
12		$\xi_3^2 \xi_4^2 - \xi_4^2 - \xi_3^2 + 1$	4
13		$\xi_3 \xi_4^3 - 3 \xi_3 \xi_4$	6
14		$\xi_4^4 - 6\xi_4^2 + 3$	24

Table 4: Two-Dimensionnal Polynomial Chaoses and their variance

$\mathbf{1}$	Ψ_0
ξ_3	Ψ_1
ξ_4	Ψ_2
ξ_3^2	$\Psi_3 + \Psi_0$
$\xi_3\xi_4$	Ψ_4
ξ_4^2	$\Psi_5 + \Psi_0$
ξ_3^3	$\Psi_6 + 3\Psi_1$
$\xi_3^2\xi_4$	$\Psi_7 + \Psi_2$
$\xi_3 \xi_4^2$	$\Psi_8 + \Psi_1$
ξ_4^3	$\Psi_9+3\Psi_2$
ξ_3^4	$\Psi_{10} + 6\Psi_3 + 3\Psi_0$
$\xi_3^3 \check{\xi}_4$	$\Psi_{11} + 3\Psi_0$
$\xi_3^2 \xi_4^2$	$\Psi_{12} + \Psi_5 + \Psi_3 + \Psi_1$
$\xi_3 \xi_4^3$	$\Psi_{13} + 3\Psi_4$
ξ_4^4	$\Psi_{14} + 6\Psi_5 + 3\Psi_0$

Table 5: Identification

Excitation characteristics The unbalance force due to an eccentric mass on a disk can be written on the degree of freedom $[v \ w \ \theta \ \psi]^\text{T}$

$$
\{\mathbf{Q}\} = m_b r_b \omega^2 e^{i\phi} [1 \quad -i \quad 0 \quad 0]^\mathbf{T} \tag{37}
$$

where m_b and r_b are the unbalance mass and the eccentricity respectively. Furthermore, ϕ defines the initial angular position. Parameters m_b and ϕ are considered as random Gaussian type quantities and are defined as

$$
m_b(\tau) = \overline{m_b}(1 + \delta_m \xi_5) \tag{38}
$$

$$
\phi(\tau) = \sigma_{\phi} \xi_6 \tag{39}
$$

with $\overline{m_b}$ and δ_m being the mean value and the variation coefficient of the unbalance mass, and σ_ϕ the standard deviation of the angular position of the force. For the reader comprehension, the angular position is illustrated in Figure 2. In addition, ξ_5 and ξ_6 are Gaussian random variables.

Figure 2: Unbalance model

Expanding $e^{i\phi}$ such as

$$
e^{i\phi} = \sum_{j=0}^{\infty} \frac{(i\phi)^j}{j!}
$$
 (40)

and substituting Equation (39) in Equation (40) leads, after truncation at a given order M, to the new expression of $e^{i\phi}$

$$
e^{i\phi} = \sum_{j=0}^{M} \frac{(i\sigma_{\phi}\xi_6)^j}{j!} \tag{41}
$$

Thus the unbalance force due to an eccentric mass on a disk can be given by

$$
\{\mathbf{Q}\} = \overline{m_b} r_b \omega^2 (1 + \delta_m \xi_5) \sum_{j=0}^{M} \frac{(i \sigma_{\phi} \xi_6)^j}{j!} [1 - i \quad 0 \quad 0]^{\mathbf{T}} \tag{42}
$$

Equation (42) can be rewritten as

$$
\{\mathbf{Q}\} = \sum_{k=0}^{1} \sum_{j=0}^{M} \{\mathbf{Q}\}_{kj} \xi_5^k \xi_6^j \tag{43}
$$

where

$$
\{\mathbf{Q}\}_{kj} = \overline{m_b} r_b \omega^2 (\delta_m)^k \frac{(i\sigma_\phi)^j}{j!} [1 \quad -i \quad 0 \quad 0]^\mathbf{T} \tag{44}
$$

Finally, the random loading {**Q**} can be expanded on the polynomial chaos basis as follow

$$
\{\mathbf{Q}\} = \sum_{j=0}^{R} \{\mathbf{Q}\}_j \Psi_j(\xi_5, \xi_6)
$$
\n(45)

where the deterministic coefficients $\{Q\}_j$ are given by using the same identification process described for the mass and gyroscopic matrices, adapted here to coefficients $\xi_5^i \xi_6^j$ and $\Psi_j(\xi_5,\xi_6)$. R is the number of polynomial chaoses given by Equation (17) for $L = 2$ and depending on the equivalent M to the $p = M + 1$ order.

Synthesis Finally, the system to be solved, given by Equation (16), is expanded on six-dimensional polynomial chaoses $\Psi_j(\xi)$ with $\xi = {\xi_1, ..., \xi_6}$, $j = 0$ to P where P is defined by Equation (17) with $L = 6$. Thus $[\mathbf{A}]_i$ is a function of $[\hat{\mathbf{M}}]_i$, $[\hat{\mathbf{C}}]_i$, $[\hat{\mathbf{G}}]_i$, $[\hat{\mathbf{K}}]_i$ and $[\hat{\mathbf{K}}_b]_i$ (see Equation (13)) which refer to a rearrangement of $[\mathbf{M}]_j, [\mathbf{C}]_j, [\mathbf{G}]_j, [\mathbf{K}_j]_j$ and $\{\mathbf{Q}\}_j$ on a six-dimensional polynomial chaos basis.

4 Numerical results

In this section, the quantification of the uncertainty effects on the response variability of the rotor under study are presented using the Polynomial Chaos Expansion method. To show lower and higher dynamic responses of the rotor system under uncertain parameters, the stochastic response of the rotor system is proposed via the mean value and the variance of the random response, also represented graphically by an envelope. The envelope of the stochastic response is constructed by calculating the maximum and the minimum of all the responses computed by the PCE approach for samples generated by the MCS method. Then, the MCS method generates n values of Hermite polynomials and consequently *n* samples of the Frequency Response Functions.

Finally, the envelope is built by considering the maximum and the minimum of all the samples.

In the following, the sections are organized as follows: firstly, the main dynamic characteristics are investigated in the deterministic case. The efficiency and robustness of the Polynomial Chaos Expansion method is then discussed for the dynamic response of the rotor system under uncertain parameters. Finally, the mean and the variance of the Frequency Response Function obtained via the PCE approach, and the envelope are compared with results obtained by using the Monte Carlo simulation.

Table 6: Sets of parameters

4.1 Deterministic case

Before discussing the effects of uncertainties on the dynamic of the rotor system, a brief summary is given of the main dynamic characteristics of the deterministic rotor system. Considering the model parameters given in Table 1, Figure 3 shows the horizontal steady-state responses of the rotor for each of its transversal nodes.

It can be seen that the horizontal displacements indicate the presence of the first, second and third forward critical speeds around 28.3 Hz, 97.2 Hz and 240 Hz, respectively. To facilitate comprehension, the first, second and third backward critical speeds do not appear on the unbalance responses due to the fact that the bearing stiffnesses are identical in the vertical and horizontal directions. Table 7 summarizes the values of the three first forward and backward critical speeds of the rotor system.

Critical speed	Value (Hz)
1st backward	27.9
1st forward	28.3
2nd backward	61.8
2nd forward	97.2
3rd backward	128.1
3rd forward	240

Table 7: Critical speeds of the rotor system

4.2 Comparisons between the Polynomial Chaos Expansion approach and Monte Carlo simulation

In this part of the paper, the results of the Monte Carlo simulation and those of the Polynomial Chaos Expansion method are compared in order to validate the efficiency of the second approach. Comparisons are given for the set of parameters defined by Case 1 in Table 6: to simulate the variation

Figure 3: Frequency Response Functions for the determinism case

of mechanical properties, the Young modulus E of the shaft and the horizontal bearing stiffness k_{1x} on the left side of the rotor system are allowed to undergo 5% variations ($\overline{E} \pm \delta_E$ and $k_{1x} \pm \delta_{k_{1x}}$).

The Monte Carlo analysis is carried out to obtain a statistical sample of the random response. As explained previously, this method requires a large number of samples to provide a reference solution. In this study, to obtain convergence of the Monte Carlo method, 1000 samples were used. For this first case, the order of chaos equals 2. A convergence study, presented soon after, will justify the choice of this truncation.

Figures 4 and 5 show the mean and the variance of the Frequency Response Function (at the node 2 in the direction x) obtained by the two methods. The two methods yield quasi-identical results for both quantities (a very low discrepancy can only be seen on peak around 100 Hz) which validates the PCE method.

Figure 6 shows the results for both the Monte Carlo simulations and the PCE method. All the Frequency Response Functions samples obtained by using the Monte Carlo simulations can be seen at node 2 in the horizontal direction, with their mean value. We can see that the mean of the Frequency Responses Function and so the envelopes built from the Monte Carlo simulations and the Polynomial Chaos Expansion are very close one to the other. It should be noted that the same samples have been used for both direct Monte Carlo method and PCE approach. Moreover, for this example, if we compare the CPU time, it appears that PCE approach is eight times faster than the direct Monte Carlo approach.

The variations of the Young modulus E of the shaft and the horizontal bearing stiffness k_{1x} can be seen to cause small changes in the critical speeds. Moreover, increases (and decreases) of the maximum amplitudes when the rotor passes through the forward critical speeds are also indicated. Finally, it can be seen that backward critical speeds (at 61.8 Hz and 128.1 Hz) can occur due to the randomness of the bearing stiffness k_{1x} which introduces dissymmetry in the rotor system.

A convergence study of the PCE with the order of chaos is performed through two choices of orders : 2 and 10. Figures 7-8 and 9-10 present the results for cases 2 and 3 (10% of the variation for the horizontal bearing stiffness k_{1x} on the right side of the rotor system. See Table 6) for the Frequency Response Functions at node 2 in the horizontal direction. As expected, the order of chaos improves modeling. However, the effect of this discrepancy between both expansions does not harm the quality of the model, especially when taking into account the increase of computation costs (which depends on the size of the problem as a function of the order of chaos) obtained subsequently. Even if the effect of damping is not on the scope of the study, it can be mentionned that decreasing the damping factor needs to increase the PCE order, especially close to the resonances. For more details, the reader is referred to the research of Dessombz [Dessombz (2000)].

Figure 4: Mean of Frequency Response Functions (Case 1); Polynomial Chaos method (red dotteddashed line); Monte Carlo Simulation (black line)

4.3 Effects of uncertainties in mechanical properties and external forces

In this part of the paper, the effects of uncertainties from stiffness properties of the rotor, geometric parameters of the disc and external forces are investigated. In the next paragraphs, only the mean

Figure 5: Variance of Frequency Response Functions (Case 1); Polynomial Chaos method (red dotteddashed line); Monte Carlo Simulation (black line)

value and the envelope of the FRF are studied in order to highlight the results more clearly. It should be noted that in the following parts all the results are given at node 2 in direction x .

4.3.1 Excitation

The effects of uncertainties in the external forcing functions are now studied. Variations on the mass unbalance and the angular position are considered. Two cases are studied (cases 4 and 5): the first and second cases deal with 1% and 5% of uncertainties for both the mass unbalance and the angular position ($m \pm \delta_m$ and $\phi \pm \delta_\phi$).

Figures 11 and 12 illustrate the mean values (using the Monte Carlo simulation and the Polynomial Chaos method) and the envelope. In this particular case, the random quantities are only located at the loading {**Q**}. Therefore, theoretically speaking, the mean of the response and the values of the critical speed obtained by the Polynomial Chaos method must be identical to the reference mean response. In Figure 11, the mean of the Polynomial Chaos approach is close to the reference mean response: the error between these two results is only due to the truncated expansion of the loading expression defined in Equation (41).

It can then be seen that the variations of the maximum amplitudes due to uncertainties for all the critical speeds are not very great. The amplitudes of the first, second and third critical speeds increase

Figure 6: Frequency Response Functions (Case 1); Mean of the FRF with Polynomial Chaos method (red dotted-dashed line); Lower and upper envelopes (red dashed line); Mean of the FRF with the Monte Carlo Simulation (black solid line); Monte Carlo samples (grey solid line)

from only 2.814 × 10⁻³m, 2.803 × 10⁻³m and 3.76 × 10⁻³m to 2.954 × 10⁻³m, 3.534 × 10⁻³m and 3.987×10^{-3} m respectively.

4.3.2 Uncertainties in disc properties

In this paragraph, variations for the properties of the disc located at the left side of the rotor system are considered. Two cases are investigated (cases 6 and 7): firstly, the density and the diameter of the disc are allowed to undergo 1% variations ($\rho \pm \delta_{\rho}$ and $D \pm \delta_{D}$). Secondly, 5% variations on the same set of parameters are introduced. As explained previously in Section 3.2, these random parameters affect the mass and gyroscopic matrices (see Equation (28)). The order of chaos has been chosen as equal to 2.

Figures 13 and 14 give the mean value of the Frequency Response Function and the envelope at node 2 in the horizontal direction. It appears that the mean values calculated by applying the Monte Carlo simulations and the Polynomial Chaos method are very similar. It should be noted that an order 2 gives accurate results in spite of the fact that the development given in Section 3.2 shows that an order 5 is needed to take all terms into account.

It can be seen that increasing uncertainties on the density and the diameter of the disc can dras-

Figure 7: Study with chaos order 2 : Mean of Frequency Response Functions (Case 2); Polynomial Chaos method (red dotted-dashed line); Monte Carlo Simulation (black line)

tically affect the values of the critical speeds and the associated maximum amplitudes. Even if the maximum variations of amplitudes are located at the critical speed, the evolutions of the rotor response far from the critical speed are significant. When comparing Figures 6 and 14, it may be concluded that the effects of uncertainties on disc properties are greater than those on shaft properties of the rotor under study.

4.3.3 Uncertainties in both mechanical properties and external forces

In order to demonstrate the efficiency and accuracy of the Polynomial Chaos procedure described above, this last part of the paper treats the cases in which uncertain quantities come from all the parameters studied previously (i.e. stiffness properties of the rotor, geometric parameters of the disc and external forces).

Numerical simulations are given by considering the variations of mechanical properties of the shaft (i.e. the Young modulus E and the horizontal bearing stiffness k_{1x}), the properties of the disc (i.e. the density ρ and the diameter D), and the excitation forces (i.e. the mass unbalance m and and the angular position ϕ), as indicated in Table 6 for cases 8 and 9. We recall that in this case the cost of calculation may be high since it is directly linked to the number of polynomials and consequently to the order of chaos and the number of random parameters. Figures 15-16 and 18-19 show the results

Figure 8: Study with chaos order 2 : Variance of Frequency Response Functions (Case 2); Polynomial Chaos method (red dotted-dashed line); Monte Carlo Simulation (black line)

for cases 8 and 9 through the mean and the variance obtained from the Polynomial Chaos and the Monte Carlo approaches. The results from both methods are in very good agreement for both the mean and for the variance. Figures 17 and 20 illustrate all the Frequency Response Function samples (at node 2 in the horizontal direction) obtained by using the Monte Carlo simulations, and the lower and upper envelopes built with the Polynomial Chaos method. It appears that increasing uncertainties affects the maximum amplitudes of the dynamic response and the value of the critical speeds. Then, as explained previously in section 4.2, the dissymmetry due to the variations in the bearing stiffness k_{1x} leads to increases in the dynamic response of the rotor system around the backward critical speeds (at 61.8Hz and 128.1Hz). Finally, it can be seen that the Polynomial Chaos method may over-estimate vibrational amplitude (see for example the dynamic response of the rotor around the third critical speed, at 240Hz). However, whatever the levels and different kinds of uncertainty (such as material, geometrical and loading characteristics) presented here, the Polynomial Chaos method agrees very well with the Monte Carlo simulation, thereby demonstrating the robustness of the method.

Figure 9: Study with chaos order 10 : Mean of Frequency Response Functions (Case 3); Polynomial Chaos method (red dotted-dashed line); Monte Carlo Simulation (black line)

5 Conclusion

This paper described a numerical procedure using the Chaos Polynomial approach to evaluate the stochastic response of a rotor system with uncertain mechanical parameters and uncertain external forces. It explained how this kind of problem can be solved with the Spectral Finite Element Method and how the random parameters can be modeled by random variables through a Karhunen-Loeve expansion. The results obtained by applying the Polynomial Chaos Expansion (PCE) procedure were compared with those evaluated by the Monte Carlo Simulation (MCS).

The stochastic response of the rotor system was proposed via the mean value and the variance of the random response and also represented graphically by an envelope. This envelope could be useful for designing rotor systems and predicting their lower and higher dynamic responses under uncertain parameters.

The efficiency and robustness of the Polynomial Chaos method were tested and validated through numerical simulations of the effects of uncertainties and orders of polynomial chaos.

Figure 10: Study with chaos order 10 : Variance of Frequency Response Functions (Case 3); Polynomial Chaos method (red dotted-dashed line); Monte Carlo Simulation (black line)

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A Model of the shaft

As illustrated in Figure 1, the nodal displacement of a beam element is defined by

$$
\delta = [v_1 \ w_1 \ \theta_1 \ \psi_1 \ v_2 \ w_2 \ \theta_2 \ \psi_2]^{\mathrm{T}}
$$
\n(46)

and for this element, the mass matrix $[\mathbf{M}^e] = [\mathbf{M}^e_1] + [\mathbf{M}^e_2]$ (summation of the translational and rotatory mass matrices), the stiffness matrix $[\textbf{K}^e]$, the gyroscopic matrix $[\textbf{G}^e]$ and the damping matrix [C^e] are expressed as

Figure 11: FRFs with randomness on the loading (Case 4); Mean of the FRF with the Polynomial Chaos method (red dotted-dashed line); Lower and upper envelopes (red dashed line); Mean of the FRF with the Monte Carlo Simulation (black solid line)

$$
\begin{bmatrix}\n\mathbf{M}_{1}^{e} = \frac{\rho SI}{420} & 156 & 0 & 0 & -22l & 54 & 0 & 0 & 13l \\
 & 156 & 22l & 0 & 0 & 54 & -13l & 0 \\
 & 4l^{2} & 0 & 0 & 13l & -3l^{2} & 0 \\
 & 4l^{2} & -13l & 0 & 0 & -3l^{2} \\
 & 156 & 0 & 0 & 22l & 0 \\
 & 156 & -22l & 0 & -4l^{2} & 0 \\
 & -4l^{2} & 0 & -4l^{2}\n\end{bmatrix}
$$
\n
$$
(47)
$$

Figure 12: FRFs with randomness on the loading (Case 5); Mean of the FRF with the Polynomial Chaos method (red dotted-dashed line); Lower and upper envelopes (red dashed line); Mean of the FRF with the Monte Carlo Simulation (black solid line)

$$
\begin{bmatrix}\n36 & 0 & 0 & -3l & -36 & 0 & 0 & -3l \\
36 & 3l & 0 & 0 & -36 & 3l & 0 \\
4l^2 & 0 & 0 & -3l & -l^2 & 0 \\
4l^2 & 3l & 0 & 0 & -l^2 \\
36 & 0 & 0 & 3l & 36 & -3l & 0 \\
36 & -3l & 0 & 4l^2 & 0 \\
4l^2 & 0 & 4l^2 & 0 & 4l^2\n\end{bmatrix}
$$
\n(48)

Figure 13: FRFs with randomness on the disc parameters (Case 6); Mean of the FRF with the Polynomial Chaos method (red dotted-dashed line); Lower and upper envelopes (red dashed line); Mean of the FRF with the Monte Carlo Simulation (black solid line)

$$
\begin{bmatrix}\n\mathbf{K}^{e} = \frac{EI}{l^3} & 0 & 0 & 6l & -12 & 0 & 0 & 6l \\
 & 12 & -6l & 0 & 0 & -12 & -6l & 0 \\
 & & 4l^2 & 0 & 0 & 6l & 2l^2 & 0 \\
 & & & 4l^2 & -6l & 0 & 0 & 2l^2 \\
 & & & 12 & 0 & 0 & -6l & 0 \\
 & & & 12 & 6l & 0 & 0 \\
 & & & 4l^2 & 0 & 4l^2 & 0 \\
 & & & & 4l^2 & 0 & 4l^2\n\end{bmatrix}
$$
\n(49)

Figure 14: FRFs with randomness on the disc parameters (Case 7); Mean of the FRF with the Polynomial Chaos method (red dotted-dashed line); Lower and upper envelopes (red dashed line); Mean of the FRF with the Monte Carlo Simulation (black solid line)

$$
\begin{bmatrix} \mathbf{G}^e \end{bmatrix} = \frac{\rho I}{15l} \begin{bmatrix} 0 & -36 & 3l & 0 & 0 & 36 & 3l & 0 \\ 0 & 0 & 3l & -36 & 0 & 0 & 3l \\ 0 & -4l^2 & 3l & 0 & 0 & l^2 \\ 0 & 0 & 3l & -l^2 & 0 \\ 0 & -36 & -3l & 0 & 0 \\ 0 & 0 & -3l & 0 & -3l \\ 0 & 0 & -4l^2 & 0 & 0 \end{bmatrix}
$$
 (50)

in which ρ and E are the density and the Young modulus of the shaft. I is the second moment of the area about any axis perpendicular to the rotor axis. S is the area of the cross section.

Figure 15: Mean of Frequency Response Functions (Case 8); Polynomial Chaos method (red dotteddashed line); Monte Carlo Simulation (black line)

B Mass and gyroscopic matrices components

$$
m_d(\tau) = \frac{1}{4}\overline{\rho}\pi h((\overline{D}^2 - d^2) + 2\delta_D \overline{D}^2 \xi_3 + \delta_\rho (\overline{D}^2 - d^2)\xi_4 + \overline{D}^2 \delta_D^2 \xi_3^2 + 2\delta_D \delta_\rho \overline{D}^2 \xi_3 \xi_4 + \delta_\rho \delta_D^2 \overline{D}^2 \xi_3^2 \xi_4)
$$
\n(51)

$$
=\sum_{j=0}^{4}\sum_{i=0}^{1}m_{dij}\xi_3^j\xi_4^i
$$
\n(52)

$$
I_p(\tau) = \frac{1}{32} \overline{\rho} \pi h ((\overline{D}^4 - d^4) + 4 \delta_D \overline{D}^4 \xi_3 + \delta_\rho (\overline{D}^4 - d^4) \xi_4 + 6 \overline{D}^4 \delta_D^2 \xi_3^2 + 4 \delta_D \delta_\rho \overline{D}^4 \xi_3 \xi_4 + 4 \delta_D^3 \overline{D}^4 \xi_3^3)
$$
\n(53)

$$
+ 6\delta_{\rho}\delta_{D}^{2}\overline{D}^{4}\xi_{3}^{2}\xi_{4} + \delta_{D}^{4}\overline{D}^{4}\xi_{3}^{4} + 4\delta_{D}^{3}\delta_{\rho}\overline{D}^{4}\xi_{3}^{3}\xi_{4} + \delta_{\rho}\delta_{D}^{4}\overline{D}^{4}\xi_{3}^{4}\xi_{4})
$$

=
$$
\sum_{j=0}^{4} \sum_{i=0}^{1} I_{p_{ij}}\xi_{3}^{j}\xi_{4}^{i}
$$
 (54)

Figure 16: Variance of Frequency Response Functions (Case 8); Polynomial Chaos method (red dotted-dashed line); Monte Carlo Simulation (black line)

$$
I_d(\tau) = \frac{1}{64} \overline{\rho} \pi h ((\overline{D}^4 - d^4) + 4 \delta_D \overline{D}^4 \xi_3 + \delta_\rho (\overline{D}^4 - d^4) \xi_4 + 6 \overline{D}^4 \delta_D^2 \xi_3^2 + 4 \delta_D \delta_\rho \overline{D}^4 \xi_3 \xi_4 + 4 \delta_D^3 \overline{D}^4 \xi_3^3
$$
\n
$$
+ 6 \delta_\rho \delta_D^2 \overline{D}^4 \xi_3^2 \xi_4 + \delta_D^4 \overline{D}^4 \xi_3^4 + 4 \delta_D^3 \delta_\rho \overline{D}^4 \xi_3^3 \xi_4 + \delta_\rho \delta_D^4 \overline{D}^4 \xi_3^4 \xi_4)
$$
\n
$$
+ \frac{1}{48} \overline{\rho} \pi h^3 ((\overline{D}^2 - d^2) + 2 \delta_D \overline{D}^2 \xi_3 + \delta_\rho (\overline{D}^2 - d^2) \xi_4 + \overline{D}^2 \delta_D^2 \xi_3^2 + 2 \delta_D \delta_\rho \overline{D}^2 \xi_3 \xi_4 + \delta_\rho \delta_D^2 \overline{D}^2 \xi_3^2 \xi_4)
$$
\n
$$
= \sum_{j=0}^4 \sum_{i=0}^1 I_{dij} \xi_3^j \xi_4^i
$$
\n(56)

Figure 17: Frequency Response Functions (Case 8); Mean of the FRF with the Polynomial Chaos method (red dotted-dashed line); Lower and upper envelopes (red dashed line); Mean of the FRF with the Monte Carlo Simulation (black solid line); Monte Carlo samples (grey solid line)

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Figure 18: Mean of Frequency Response Functions (Case 9); Polynomial Chaos method (red dotteddashed line); Monte Carlo Simulation (black line)

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Figure 19: Variance of Frequency Response Functions (Case 9); Polynomial Chaos method (red dotted-dashed line); Monte Carlo Simulation (black line)

Figure 20: Frequency Response Functions (Case 9); Mean of the FRF with the Polynomial Chaos method (red dotted-dashed line); Lower and upper envelopes (red dashed line); Mean of the FRF with the Monte Carlo Simulation (black solid line); Monte Carlo samples (grey solid line)