Polynomial chaos expansion for acoustic propagation

Séminaire des doctorants du CMLA

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CONTEXT

► The verification regime of the Comprehensive Nuclear-Test-Ban Treaty (CTBT) is designed to detect any nuclear explosion conducted on Earth – underground, underwater or in the atmosphere.

► Infrasound monitoring is one of the four technologies used by the International Monitoring System (IMS) to verify compliance with the CTBT.





► Infrasound has the ability to cover long distances with little dissipation.

▶ Infrasound signals can be severly distorted by the propagation in the atmosphere.

PLANETARY BOUNDARY LAYER



NORMAL MODES (1/2)



 $p(t) = \mathcal{F}^{-1}[G(\omega)s(\omega)](t)$

NORMAL MODES (2/2)

Signal reconstructed with only one mode:

$$p_i(t) = \mathcal{F}^{-1}[G_i(\omega)s(\omega)](t)$$



RANDOM JET



Impact of the uncertainties on the medium on the acoustic signal received at the ground?

> Metamodel of **the eigenpairs of the propagation operator** able to generate such signals.

POLYNOMIAL CHAOS DECOMPOSITION (1/2)

Build a metamodel of $X = \mathcal{M}(\boldsymbol{\xi})$ where $\boldsymbol{\xi} \sim \mathcal{N}(0, \mathbb{I}_n)$:



POLYNOMIAL CHAOS

Polynomial chaos decomposition:

 $\blacksquare (H_j)_{j \in J} \text{ set of orthonormal polynomials for } \langle f, g \rangle = \mathbb{E}[fg]$

$$X(\boldsymbol{\xi}) = \sum_{j \in J} a_j H_j(\boldsymbol{\xi}) \text{ where } a_j = \langle X, H_j \rangle$$

taking polynomials up to degree d, $|J| = \frac{(n+d)!}{n!d!}$

POLYNOMIAL CHAOS DECOMPOSITION (2/2)

Coefficients $(a_j)_{j \in J}$ can be computed:

with Monte-Carlo using a sample *S*:



with a quadrature \mathcal{Q} :



GPC DECOMPOSITION OF THE ACOUSTIC MODES

gPC decomposition of each eigenvalue and eigenvector at the ground

$$\widehat{k_i}(\omega, \boldsymbol{\xi}) = \sum_{j \in J} a_j^{k_i}(\omega) H_j(\boldsymbol{\xi}) \text{ and } \widehat{\Psi_i}(\omega, \boldsymbol{\xi}) = \sum_{j \in J} a_j^{\Psi_i}(\omega) H_j(\boldsymbol{\xi})$$

Reconstruction of the Green function

$$\widehat{G}(\omega,\boldsymbol{\xi}) = \sum_{i=1}^{N} \widehat{G}_{i}(\omega,\boldsymbol{\xi}) = \sum_{i=1}^{N} \alpha \frac{\widehat{\Psi_{i}}(\omega,\boldsymbol{\xi})^{2}}{\sqrt{\widehat{k_{i}}(\omega,\boldsymbol{\xi})R}} e^{i\widehat{k_{i}}(\omega,\boldsymbol{\xi})R}$$

Simulation of signals using the metamodel:

$$p(t, \boldsymbol{\xi}) = \mathcal{F}^{-1}[\widehat{G}(\omega, \boldsymbol{\xi})s(\omega)](t)$$

CONVERGENCE

Variance can be used to control the convergence:



Statistics on the signals can be compared with those obtained by Monte-Carlo simulations:



SENSITIVITY ANALYSIS

Sobol index S_j(k_i) gives a measure of the sensibility of mode k_i to the parameters ξ_j:

$$S_j(k_i) = \frac{Var(\mathbb{E}[k_i|\xi_j])}{Var(k_i)}$$

The gPC decomposition allows a quick computation of the Sobol indices: $S_j(k_i) = \sum_{j \in J'} (a_j^{k_i})^2 \text{ where } J' = \{k \in J^* | \exists Q \in \mathbb{R}[X], P_k(\boldsymbol{\xi}) = Q(\xi_j)\}$



Conclusion:

- **gPC** representation of the acoustic modes
- Metamodel able to give statistics on the signals
- Sensitivity analysis on the modes

Perspectives:

- Use model reduction to select modes with high acoustical contribution and great sensitivity to the uncertainties.
- Develop an eigenvalue tracking method to generalize to a realistic atmosphere.