





#### Motivation

Reliable Simulations of QPR
Parametrized model of QPR
Stochastic Forward Problem
Pseudo-spectral Approach

Result for MO robust shape optimization of QPR Formulation of MO robust optimization



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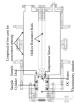


# Sources of uncertainty in accelerator physics

Reliable simulations of the superconducting radio-frequency cavity:

- input data & code uncertainties : measurement errors, error propagation
- modeling uncertainties : calibration and validation
- manufacturing uncertainties: ultrasonic bath, buffered chemical polishing, etc.
  - roughness of the superconducting surfaces
  - · affect the material and geometrical parameters

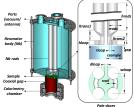






#### **QPR**

#### Short characteristics of a device



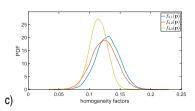
- special dedicated device used for the measurement of the surface resistance of superconducting samples at temperatures of 1.8 K to more than 20 K
- RF fields of up to 120 mT and operating frequencies of 433 MHz, 866 MHz and 1.3 GHz
- composed of a pillbox-like cavity containing four-vertically placed hollow rods
- quadrupole-like magnetic field is excited on the superconducting sample
- measurement data and expected losses on the sample allows for measuring the surface resistance of the sample with aid of calorimetric method

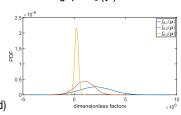


# Consequences of Uncertainties

The measurement procedure is affected by various sources of uncertainties

- associated with the resolution of electronic equipment, geometrical deviations
  of a cavity design, and the accuracy of numerical simulations
- various figures of merit : a) operating frequencies  $-f_0(\mathbf{p})$ , b) focus factor  $-f_1(\mathbf{p})$ , c) homogeneity of the magnetic field distribution on the sample  $-f_2(\mathbf{p})$ , d) penetration of the magnetic field into the coaxial gap  $-f_3(\mathbf{p})$

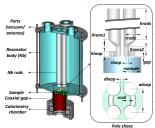






## Parametrized model of QPR

A three-dimensional symmetric model (3D)



- as optimization of uncertain design parameters : Q = 5,  $\mathbf{p} := (p_1, p_2, p_3, p_4, p_5)^{\top} = (gap, rrods, hloop, rloop, wloop)^{\top}$ .
- geometrical imperfections related to the manufacturing of the QPR of order 50 100 [ $\mu m$ ] are mimicked by the Gaussian distribution as follows

$$p_{\mathbf{q}}(\xi) = \overline{p}_{\mathbf{q}}(1 + \delta_{\mathbf{q}} \cdot \xi_{\mathbf{q}}), \quad q = 1, \dots, Q,$$

 $\delta_{\bf q}$  allows for controlling the magnitude of the perturbation such that  $\sigma_{\bf q}:=\delta_{\bf q}\cdot\overline{\rho}_{\bf q}=0.05\,[{\rm mm}]$ 



# Stochastic Maxwell's Eigenproblem

Eigenpairs ( $E(p), \lambda(p)$ ):

$$\begin{split} -\nabla \times \left( \nu \, \nabla \times \mathbf{E}(\mathbf{x}, \mathbf{p}) \right) + \lambda(\mathbf{p}) \, \epsilon \, \mathbf{E}(\mathbf{x}, \mathbf{p}) &= 0, & \text{in } D, \\ \mathbf{n} \times \mathbf{E}(\mathbf{x}, \mathbf{p}) &= 0, & \text{on } \partial D_{\mathrm{P}}, \\ \mathbf{n} \times \left( \nu \nabla \times \mathbf{E}(\mathbf{x}, \mathbf{p}) \right) &= 0, & \text{on } \partial D_{\mathrm{N}} \end{split}$$

for  $\mathbf{x} \in D \subset \mathbb{R}^3$ ,  $\partial D = \partial D_\mathrm{P} \cup \partial D_\mathrm{N}$ ,  $\mathbf{p} = (p_1, \dots, p_Q)^\top \in \Pi \subset \mathbb{R}^Q$ **E**: phasor of electric field,  $\lambda = \frac{\omega^2}{c^2}$ : eigenfrequency,  $\omega$ : angular frequency c: speed of light,  $\nu$ : magnetic reluctivity,  $\epsilon$ : electric permittivity

Discretization: finite element method

(triangular mesh, piecewise linear functions)



## UQ for Stochastic Forward Problem

Stochastic variables  $(\Omega, \Sigma, \mu)$ :  $\mathbf{p}(\xi) = (p_1(\xi), \dots, p_Q(\xi))$ ,  $\mathbf{p} : \Omega \to \Pi$ , independent, Gaussian, uniform, beta, etc.

Polynomial Chaos Expansion : a finite second moment of  $f: [\lambda_0, \lambda_{\mathrm{end}}]$  :

$$f(\lambda, \mathbf{p}(\xi)) \doteq \sum_{i=0}^{N} v_i(\lambda) \phi_i(\mathbf{p}(\xi))$$

Based on calculations of a model at each quadrature points  $\underline{\textbf{p}}^{(1)},\dots,\underline{\textbf{p}}^{(K)}\in\Pi$  :

$$\mathbb{E}\left[f\left(\lambda,\;\mathbf{p}\right)\right] \;=\; v_0(\lambda), \quad \mathsf{Var}\left[f\left(\lambda,\;\mathbf{p}\right)\right] \;=\; \sum_{i=1}^N \left|v_i(\lambda)\right|^2$$

by using a multi-dimensional quadrature rule with weights  $w^{(1)}, \dots, w^{(K)} \in \mathbb{R}$ :

$$v_i(\lambda) := \langle f(\lambda, \mathbf{p}), \phi_i(\mathbf{p}) \rangle \doteq \sum_{k=1}^K w_k f(\lambda, \mathbf{p}^{(k)}) \phi_i(\mathbf{p}^{(k)})$$



# Sensitivity analysis

# Local sensitivity:

$$\left. \frac{\partial f}{\partial p_j} \right|_{p_j = \overline{p}_j} = \sum_{i=0}^N v_i \frac{\partial \phi_i}{\partial p_j} \frac{\partial \mathbf{p}}{\partial \xi_j}, \quad j = 1, \dots, Q.$$

## Variance-based sensitivity:

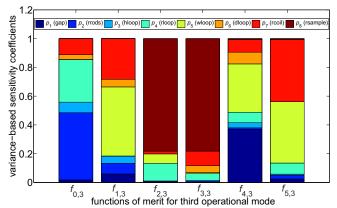
$$S_j = rac{\mathsf{V}_j}{\mathsf{Var}(f)} \quad ext{with} \quad \mathsf{V}_j := \sum_{i \in I_i} |v_i|^2, \quad j = 1, \dots, Q,$$

 $I_j$ : sets  $I_j := \{j \in \mathbb{N} : \phi_j(\mathbf{p}) \text{ is not constant in } p_j\}$  Var(f): the total variance  $0 < S_i < 1$ : upper and lower bounds



### UQ result for the end cell of QPR

Result for the variance based sensitivity analysis





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# Robust MO shape optimization

Random variables :  $\boldsymbol{\xi} = (\xi_1, \xi_2, \xi_3, \xi_4, \xi_5)$ Random dependent functionals :

$$[f_1(\mathbf{p}), f_2(\mathbf{p}), f_3(\mathbf{p})] = \left[\frac{1}{2U} \int_{\Omega_{\mathbf{S}}} \|\mathbf{H}(\mathbf{p})\|^2 d\mathbf{x}, \frac{\int_{\Omega_{\mathbf{S}}} \|\mathbf{H}(\mathbf{p})\|^2 d\mathbf{x}}{|\Omega_{\mathbf{S}}| \max_{\mathbf{x} \in \Omega_{\mathbf{S}}} (\|\mathbf{H}(\mathbf{p})\|^2)}, \frac{\int_{\Omega_{\mathbf{S}}} \|\mathbf{H}(\mathbf{p})\|^2 d\mathbf{x}}{\int_{\Omega_{\mathbf{F}}} \|\mathbf{H}(\mathbf{p})\|^2 d\mathbf{x}}\right]$$

Functionals for robust optimization:

$$\begin{split} &\inf_{\overline{\mathbf{p}} \in \mathbb{R}^{Q}} \left[ \mathbb{E}(f_{1}), \mathbb{E}(f_{2}), \mathbb{E}(f_{3}) \right]^{\top} \\ &\text{S.t. } \nabla \times \left( \nu \, \nabla \times \mathbf{E}(\mathbf{p}, \cdot) \right) - \lambda(\mathbf{p}, \cdot) \, \epsilon \, \mathbf{E}(\mathbf{p}, \cdot) = 0, \\ &\rho_{L_{\boldsymbol{q}}} \leq \overline{\rho}_{\boldsymbol{q}} \leq \rho \nu_{\boldsymbol{q}}, \text{ for } q = 1, \dots, Q \end{split}$$

Approximation of probabilistic integrals :

Stroud-3 formula (10 nodes)



## Parameters of stochastic simulation

## Variations of geometrical parameters:

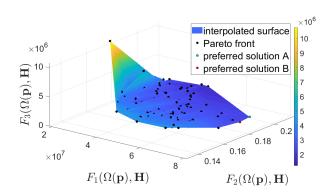
 $\longrightarrow$  modeled by Gauss distribution

# Random variations of parameters:

- independent normal random variables :  $\xi_1, \xi_2, \xi_3, \xi_4, \xi_5$
- the magnitude of perturbation :  $\sigma_q := \delta_q \cdot \overline{p}_q = 0.05 \, [\mathrm{mm}]$



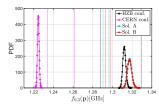
# Robust MO Shape Optimization : Pareto front VBS-MO shape optimization

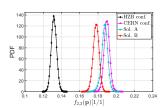


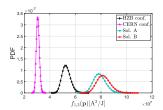


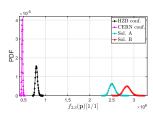
# Robust MO Shape Optimization: Pareto front

#### Probabilistic density functions











# Robust MO Shape Optimization : shapes of QPR VBS-MO shape optimization

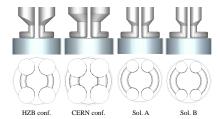


TABLE V. Results for the MO optimization - parameter domain a Name  $\Omega_{HZB}^{*}(\overline{\mathbf{p}})$  $\Omega^*_{CERN}(\overline{\mathbf{p}})$  $\Omega_{\Lambda}^{*}(\overline{\mathbf{p}})$  $\Omega_{B}^{*}(\overline{\mathbf{p}})$  $p_1$  (gap) [mm]0.50 0.700.58 0.55  $p_2$  (rrods) [mm]13.00 15.00 9.76 9.14  $p_3$  (hloop) [mm]10.00 10.00 9.72 9.64  $p_4$  (rloop) 5.92 5.56 [mm]5.00 8.00 43.79 43.53 p<sub>5</sub> (wloop) [mm]44.00 40.93  $p_6$  (dloop) [mm]6.00 5.00 4.00 4.00 p<sub>7</sub> (rcoil) [mm]22,408 23.00 25.00 25.00  $p_8$  (rsample) [mm] 37.5037.50 35.0 35.00



# Robust MO Shape Optimization: summary

TABLE VI. Results of the MO optimization for the first mode – objective space [4]

Means/Configurations	$\Omega^*_{\mathrm{HZB}}(\overline{\mathbf{p}})$	$\Omega^*_{CERN}(\overline{\mathbf{p}})$	[%]	$\Omega_{A}^{*}(\overline{\mathbf{p}})$	[%]	$\Omega_B^*(\overline{\mathbf{p}})$	[%]
$F_1(\Omega^*(\overline{\mathbf{p}}), \cdot) [M A^2/J]$	50.07	32.15	-36.55	56.31	11.13	58.47	15.39
$F_2(\Omega^*(\overline{\mathbf{p}}), \cdot)$ [1/1]	0.155	0.218	41.15	0.227	48.84	0.216	39.70
$F_3(\Omega^*(\overline{\mathbf{p}}), \cdot)$ [M 1/1]	1.668	0.890	-46.64	3.941	136.3	4.421	165.1
$F_4(\Omega^*(\overline{\mathbf{p}}), \cdot)$ [1/1]	0.910	0.906	-0.43	0.901	-1.01	0.905	-0.62
$F_5(\Omega^*(\overline{\mathbf{p}}), \cdot) [\text{mT/(MV/m)}]$	7.888	5.250	-32.93	4.824	-38.84	4.940	-37.38
$F_0(\Omega^*(\overline{\mathbf{p}}), \cdot)$ [GHz]	0.429	0.398	-7.21	0.439	2.21	0.439	2.23

<sup>&</sup>lt;sup>a</sup> The columns with percentage [%] indicate a ratio (increase +/decrease -) of optimized configurations to  $\Omega^*_{HZB}(\overline{\mathbf{p}})$ .

TABLE VIII. Results of the MO optimization for the third mode - objective space [1]

Means/Configurations	$\Omega^*_{\mathrm{HZB}}(\overline{\mathbf{p}})$	$\Omega^*_{CERN}(\overline{\mathbf{p}})$	[%]	$\Omega_A^*(\overline{\mathbf{p}})$	[%]	$\Omega_B^*(\overline{\mathbf{p}})$	[%]				
$F_1(\Omega^*(\overline{\mathbf{p}}), \cdot) [M A^2/J]$	52.28	30.63	-42.05	78.98	49.43	82.04	55.21				
$F_2(\Omega^*(\overline{\mathbf{p}}), \cdot)$ [1/1]	0.132	0.19	44.00	0.187	42.09	0.178	35.0				
$F_3(\Omega^*(\overline{\mathbf{p}}), \cdot)$ [M 1/1]	0.791	0.467	-40.89	2.501	217.4	2.846	259.9				
$F_4(\Omega^*(\overline{\mathbf{p}}), \cdot)$ [1/1]	0.914	0.917	0.3	0.907	-0.81	0.897	-1.94				
$F_5(\Omega^*(\overline{\mathbf{p}}), \cdot) \text{ [mT/(MV/m)]}$	5.048	5.411	7.19	4.736	-6.18	4.685	-7.19				
$F_0(\Omega^*(\overline{\mathbf{p}}), \cdot)$ [GHz]	1.312	1.225	-6.67	1.317	0.41	1.317	0.41				

<sup>&</sup>lt;sup>a</sup> The columns with percentage [%] indicate a ratio (increase +/decrease -) of optimized configurations to  $\Omega^*_{HZR}(\mathbf{p})$ .



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## Conclusions and further research

#### Conclusions:

- VBS robust MO shape optimization problem of QPR under uncertainties
- optimized configuration of QPR allows for increasing the focusing factor of the third mode by 50-57% and 158-168% compared to the HZB and CERN designs,
- → better resolution of the surface resistance in different freq<sub>i</sub>
- the dimensionless factor of freq<sub>3</sub> is more than twice bigger than for the HZB and CERN configuration
- it helps to decrease the measurement bias for the third mode in HZB and CERN designs

#### Further research directions:



Thank you for your attention