

An iterative tolerance analysis procedure to deal with linearized behavior models

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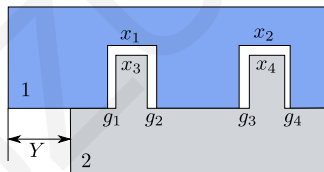


- 1 Introduction : tolerance analysis problem formulation
- 2 Linearization procedure
- 3 Solution methods
 - Monte Carlo simulation
 - Iterative algorithm
- 4 Industrial application

Formalization

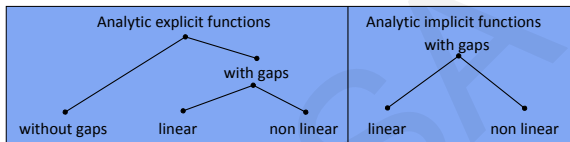
Geometrical model

- Geometrical deviations x : Random variables whose distributions are known.
- Gaps g : Free variables depending on the configuration of the mechanism.



Formalization

Behavior model



$$Y = f(x)$$

$$C_i(Y) \leq 0$$

$$C_f(Y) \leq 0$$

$$Y = f(x, g)$$

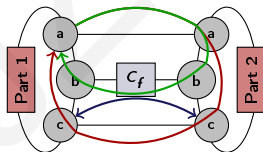
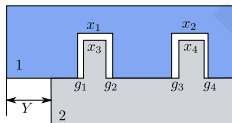
$$C_i(Y) \leq 0$$

$$C_f(Y) \leq 0$$

$$C_c(x, g) = 0$$

$$C_i(x, g) \leq 0; C_{i^*}(x, g) = 0$$

$$C_f(x, g) \leq 0$$



Compatibility equations

$$C_c(x, g) = 0$$

Interface constraints

$$C_i(x, g) \leq 0 \text{ and } C_{i^*}(x, g) = 0$$

Functional condition

$$C_f(x, g) \leq 0$$

Formalization

Quantifiers based conditions

Assembly condition

There exists an admissible gap configuration of the mechanism such that the assembly requirement (interface constraints) and the compatibility equations are respected.

$$\exists \mathbf{g} \in \mathbb{R}^m \{ C_c(\mathbf{x}, \mathbf{g}) = 0 \cap C_i(\mathbf{x}, \mathbf{g}) \leq 0 \cap C_{i^*}(\mathbf{x}, \mathbf{g}) = 0 \}$$

Use of an optimization algorithm to check the existence of an admissible configuration

$$R(\mathbf{x}) = \min_{\mathbf{g} \in \mathbb{R}^m} C_i^1(\mathbf{x}, \mathbf{g})$$

Subject to

$$\begin{aligned} C_c(\mathbf{x}, \mathbf{g}) &= 0 \\ C_i(\mathbf{x}, \mathbf{g}) &\leq 0 \\ C_{i^*}(\mathbf{x}, \mathbf{g}) &= 0 \end{aligned}$$

- $R(\mathbf{x})$ has a solution means that there exists a configuration of gaps satisfying all constraints.
- $R(\mathbf{x})$ has no solution means the mechanism cannot be assembled.

$$P_{fa} = \text{Prob} [\nexists R(\mathbf{X})]$$

Optimization algorithm

Use of an optimization algorithm to check the existence of an admissible configuration

$$\begin{aligned} R(\mathbf{x}) &= \min_{\mathbf{g} \in \mathbb{R}^m} C_i^1(\mathbf{x}, \mathbf{g}) \\ \text{Subject to } & C_c(\mathbf{x}, \mathbf{g}) = 0 \\ & C_i(\mathbf{x}, \mathbf{g}) \leq 0 \\ & C_{i^*}(\mathbf{x}, \mathbf{g}) = 0 \end{aligned}$$

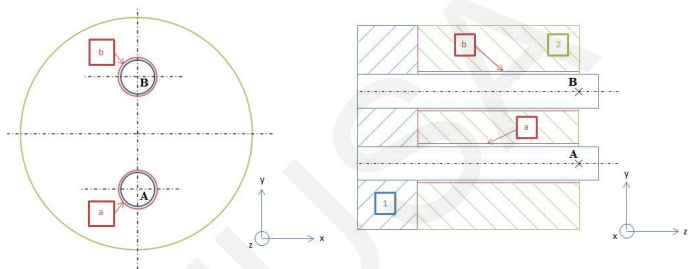
Difficulty :

- Optimization step may be complex when taking into account non linear constraints.

Considered solution :

- Linearization of non linear equations.

Example of mechanism



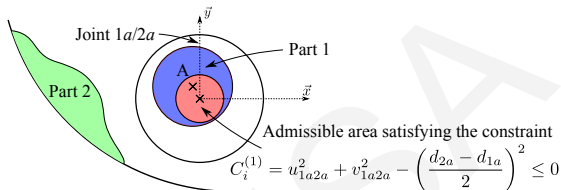
Interface constraints to be linearized

Both cylindrical joints on surfaces a and b : gap on each side must not exceed the radius difference \Rightarrow quadratic inequations :

$$u_{1a2a}^2 + v_{1a2a}^2 - \left(\frac{d_{1a} - d_{2a}}{2} \right)^2 \leq 0$$

$$u_{1b2b}^2 + v_{1b2b}^2 - \left(\frac{d_{1b} - d_{2b}}{2} \right)^2 \leq 0$$

Linearization procedure



First order linearization to point $P_k(\Delta R \cos \theta_k, \Delta R \sin \theta_k)$

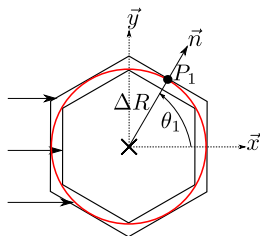
Two strategies :

- Inner polygon
- Outer polygon

Type 2: outer polygon

Type 1: inner polygon

Real admissible area



$$u_{1a2a} \cos \theta_k + v_{1a2a} \sin \theta_k - \left(\frac{d_{1a} - d_{2a}}{2} \right) - r_s \leq 0$$

with $\theta_k = \frac{2k\pi}{N_d}$ and $k = 1, \dots, N_d$

Solution method based on Monte Carlo simulation

- 1 Random sampling of N geometrical deviations $\mathbf{X} = \{X_1(\omega), \dots, X_n(\omega)\}$,
- 2 Finding an admissible configuration of gaps satisfying the behavior model using the optimization algorithm,
- 3 Estimating the probability of failure : $\tilde{P}_f = \frac{1}{N} \sum_{i=1}^N I_D(\mathbf{x}^{(i)})$, where $I_D(\mathbf{X})$ is the indicator function : $I_{D_{fa}}(\mathbf{X}) = \begin{cases} 1 & \text{if no solution can be provided} \\ 0 & \text{if a solution is found} \end{cases}$

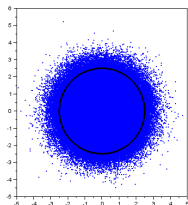
Difficulty :

- No information about the required number of linearizations N_d to get accurate enough result.
- Performing several Monte Carlo simulations is not conceivable due to prohibitive computing time.

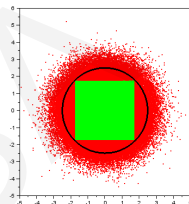
Considered solution :

- Use of an iterative algorithm.

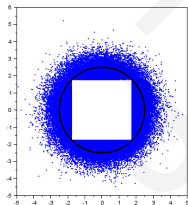
Illustration of the iterative algorithm



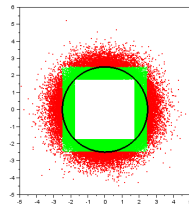
Step 1 : Define Monte Carlo population
 $N = 500000$



Step 2 : full Monte Carlo with inner strategy
 $N_d = 4$

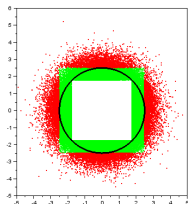


Step 3 : save non assembly points $N = 74276$

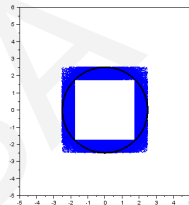


Step 4 : Monte Carlo with outer strategy
 $N_d = 4$

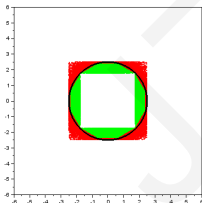
Illustration of the iterative algorithm



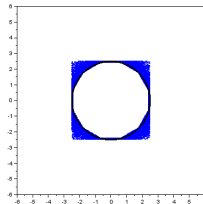
Step 4 : Monte Carlo with outer strategy
 $N_d = 4$



Step 5 : save uncertain points $N = 61858$

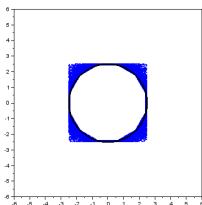


Step 6 : Monte Carlo with inner strategy
 $N_d = 12$

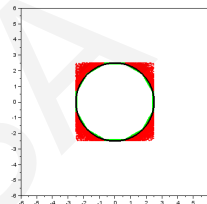


Step 7 : save non assembly points $N = 13052$

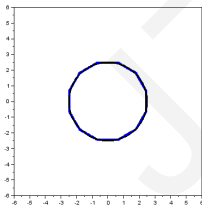
Illustration of the iterative algorithm



Step 7 : save non assembly points $N = 13052$



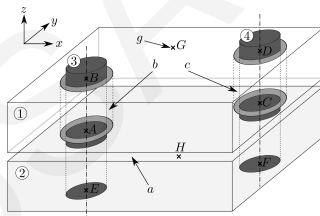
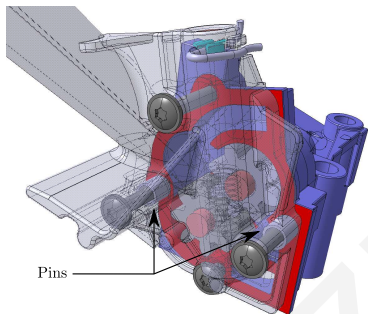
Step 8 : Monte Carlo with outer strategy
 $N_d = 12$



Step 9 : save uncertain points $N = 4949$

- The procedure stops when a stopping criterion $rCI = \frac{P_{inner} - P_{outer}}{P_{inner}}$ is small enough
- The number of required optimizations decreases at each step
- The number of calls to the most expensive models is smaller

Industrial application : planar joint with two pins



Characteristics

- 38 geometrical deviations (random variables) \mathbf{x}
- 15 gap variables \mathbf{g}
- 12 compatibility equations $C_c(\mathbf{x}, \mathbf{g}) = 0$
- $4 \times N_d$ quadratic interface constraints $C_i(\mathbf{x}, \mathbf{g}) \leq 0$

Industrial application : numerical comparison

- The number of samples (= optimization calls) of the Monte Carlo simulation is set to $14e6$, the coefficient of variation is about 5%.
- The iterative procedure starts with $N_d = 4$ and stops at $N_d = 108$.
- The stopping criterion is equal to $rCI = 1\%$.

Monte Carlo simulation

N_d	$P_{inner} \times 10^{-5}$	$P_{outer} \times 10^{-5}$	Time for one optim.	Total time
4	45.9	0.49	$1.1e^{-3}$ s	4.1 h
12	4.94	2.99	$1.8e^{-3}$ s	7.1 h
36	4.34	4.06	$5e^{-3}$ s	19.5 h
108	4.19	4.15	$22.5e^{-3}$ s	87.7 h

Iterative procedure

N_d	$P_{inner} \times 10^{-5}$	$P_{outer} \times 10^{-5}$	Optim. calls	Time
4	45.9	0.49	$14e6 + 6426$	4.16 h
12	4.94	2.99	$6357 + 622$	
36	4.34	4.06	$273 + 189$	
108	4.19	4.15	$38 + 19$	

Thank you for your attention

