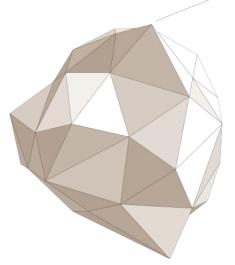
Comsol Multiphysics Simulation Integrated into Genetic Optimization

INNOVATION MAKERS



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OUTLINE



- Context & Objectives
- > Thermoelectric Micro-Generators
- Genetic Algorithm
- Optimization Tool
- > Results
- Conclusion





CONTEXT & OBJECTIVES



CONTEXT

industrial research program focused on the development of **thermoelectric micro-generators** (µTEGs) based on **innovative materials**

GOAL

innovative tool of modeling, simulation and optimization

- √ system performances improvement
- √ huge numbers of parameters and constraints
- ✓ wide range of applicability

APPROACH

interaction between Comsol Multiphysics and MathWorks Matlab



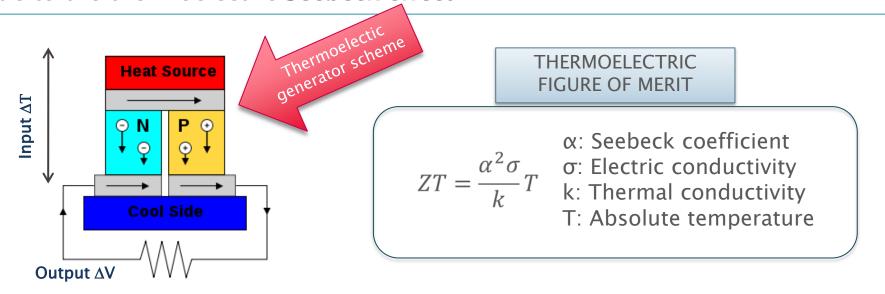


THERMOELECTRIC MICRO-GENERATOR - Description



DEFINITION

A thermoelectric micro-generator (μ TEG) is a solid state device able to generate an electric potential when it is exposed to a temperature gradient due to the thermoelectric **Seebeck effect**

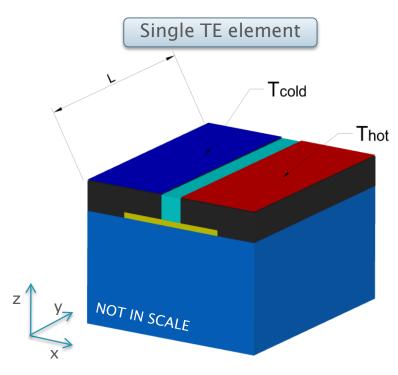


The possibility of using $\mu TEGs$ to convert waste heat into electricity has recently regained interest as a consequence of the discovery of high ZT values in a certain range of materials, but in order to fabricate an **high efficiency \mu TEG**, it is necessary to design the **optimal geometry** of the μTEG 's element!



THERMOELECTRIC MICRO-GENERATOR - Geometry



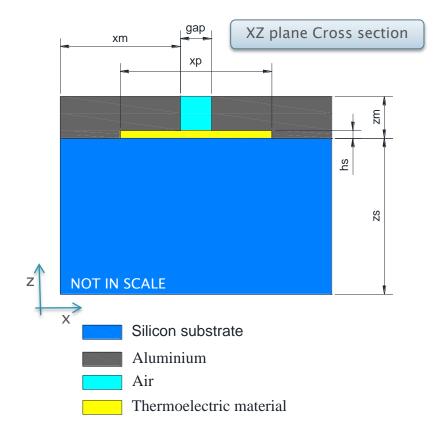


Heat exchange surfaces:

T hot side = 493.15K T cold side = 293.15K

Electric exchange surfaces:

Cold side: Potential ground Hot side: Variable potential



Variable	PROJECT VALUE
xm	55 μm
xp	7 μm
gap	4 μm
zm	6 µm

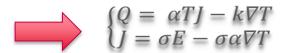


THERMOELECTRIC MICRO-GENERATOR - Physical Model



The main physical quantities are

- o heat flux (Q)
- electric current flux (J)



Governing equation for thermoelectric effect are:

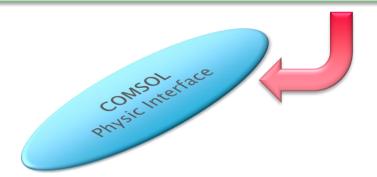
- Heat energy conservation
- Electric current balance

$$\nabla \cdot Q = J \cdot E$$

$$\nabla \cdot J = 0$$

$$\nabla \cdot [\alpha T(-\sigma \nabla V - \sigma \alpha \nabla T) - k \nabla T] = (-\sigma \nabla V - \sigma \alpha \nabla T) \cdot (-\nabla V)$$
$$\nabla \cdot (-\sigma \nabla V - \sigma \alpha \nabla T) = 0$$





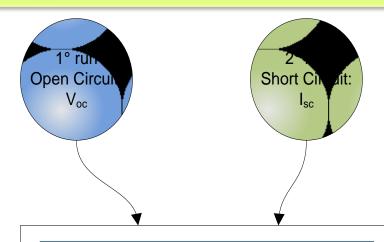


THERMOELECTRIC MICRO-GENERATOR - Comsol Model



The single TE element can be schematized with an equivalent electric circuit.

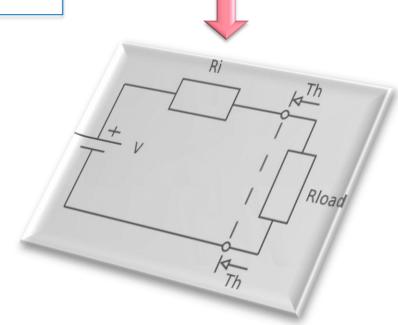
Calculation method



Post processing electric power evalutation

$$P_{load} = \frac{V^2}{4R_i^2} = \frac{V_{oc}I_{sc}}{4}$$

Max power transfert



V: Seebeck generated voltage

R_i: Internal resistance of the TE

generator

R_{load}: Load resistance



GENETIC ALGORITHM - Problem Design (1)



OPTIMIZATION PROBLEM FORMULATION

max f(x) *objective function*

such that:

 $c(x) \leq 0$

 $A \bullet X \leq b$

 $1 \leq x \leq u$

non- linear constraints

linear constraints

lower and upper bounds



FIRST CHOICE OBJECTIVE FUNCTION

Electric Power Density = Electric Power/Area (Watt/cm²)

In manufacturing processes

- √ maximize the electrical power
- ✓ minimize the area in order to reduce the probability of defect occurrence at wafer level



GENETIC ALGORITHM – Problem Design (2)



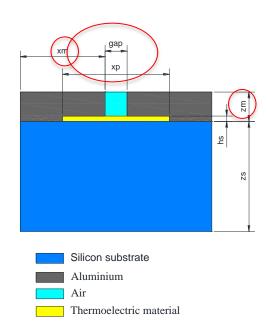


SECOND CHOICE VARIABLES TO OPTIMIZE

Upper & Lower Bound Definition

VARIABLES TO OPTIMIZE				
NAME	LOWER BOUND	UPPER BOUND		
xm	10 μm	60 µm		
xp	б µт	10 μm		
gap	4 μm	5 μm		
zm	1 μm	б µт		

Given by specialists





THIRD CHOICE CONSTRAINTS DEFINITION

Linear Constraint

$$xp - gap > 2\mu m$$

Technical constraint: lithography resolution



GENETIC ALGORITHM - Optimization approach



GENETIC ALGORITHM: <u>heuristic</u> <u>search</u> that simulates the process of

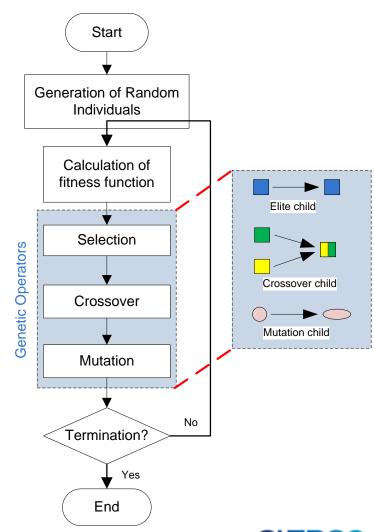
natural selection

The algorithm starts by creating a random initial feasible population; then it creates a sequence of new populations.

At each step:

- It scores each individual of the current population by computing its fitness value
- It selects members, called parents, based on their fitness, in order to create the next population.
- It creates the new population from parents by means of selection, crossover and mutation

It stops when stopping criteria is met: average cumulative change over generations

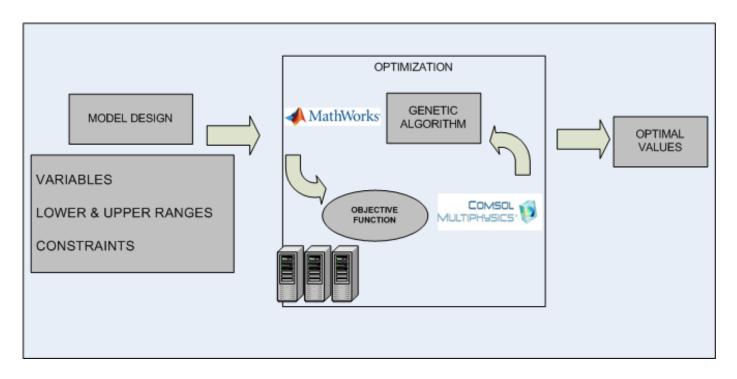




OPTIMIZATION TOOL



- Genetic Algorithm is implemented in MathWorks MatLab
- ➤ Electrical Power Density is the fitness function and its evaluation is performed by Comsol Multiphysics



At each generation, for each individual processed by genetic algorithm, MatLab sets new values into Comsol model's parameters and invokes Comsol simulations to evaluate fitness function



NUMERICAL RESULTS (1)



Genetic Algorithm finds optimal values for each variable:

Variable	PROJECT VALUE	OPTIMIZED VALUE	
xm	55 μm	10 μm	
хр	7 μm Optimi	7.36 µm	
gap	4 μm	4 μm	
zm	6 μm	2.12 μm	

Optimal values maximize electrical power density:

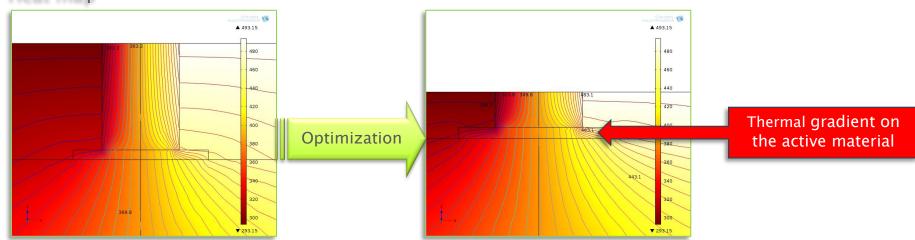
FEATURE	PROJECT VALUE	OPTIMIZED VALUE
Horizontal area (S)	$2280~\mu m^2$	$480~\mu m^2$
Electric Potential (V)	57.80 mV	69.92 mV
Electrical Current (I)	5.75 mA	7.10 mA
Heat flux	435.18 mW	274.33 mW
Electric Power generated (Pload)	0.08 mW	0.12 mW
Electric Power Density (Pd)	3.64 W/cm ²	25.86 W/cm ²



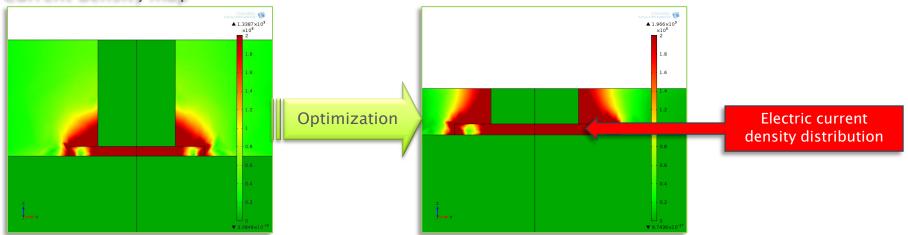
RESULTS (2)







Current density map





CONCLUSION



- ➤ This work aims at demonstrating the power of interaction of multiphysics simulation with the optimization approach of genetic algorithms
- > Numerical results must be considered demonstrative
- >The tool greatly improves μTEG's performances
- More generally, this tool can be a precious contribution to device designers, especially when device models have many variables to take in account and many complex geometrical or technical constraints.



INNOVATION MAKERS

