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Cooling channel optimization for injection molding

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Abstract. In injection molding process, heat transfer during the cooling step plays an important role. This step has a great influence on the quality of the final parts that are produced, as well as on the molding cycle time. We introduce an optimization procedure in order to locate automatically the cooling channels in 3D injection molds. The temperature distribution is computed using the Boundary Elements Method (BEM) that allows reducing the computation space from 3D to 2D, (avoiding full 3D remeshing). In our study, BEM is used to solve the stationary heat conduction problem. The BEM heat transfer solver is coupled with the non-linear optimization algorithm SQP (Sequential Quadratic Programming). The SQP algorithm permits to calculate the best set of cooling parameters, for a given cost function. For example, one such cost function involves minimizing the temperature variations at the interface between the mold cavity and the polymer. We present preliminary 3D computational results.

Keywords: BEM, heat transfer, injection molding, mold-cooling simulation, optimization, SQP. PACS: 02.70.Pt

INTRODUCTION

Numerical simulations are more and more used for designing injection molds. The location of the cooling channels is a major element in the design of the mold because the cooling time can represents up to 70% of the injection cycle. Rapid prototyping processes such as layered design or selective laser sintering start to be used to manufacture injection molds. The advantage of these assembling processes is the possibility to obtain almost any desired shape of mold geometry. However; it is then more difficult to locate and to shape the cooling channels. Numerical simulation can be used to perform automatic optimization of the position and the shape of the cooling channels.

Simulation of Heat Transfer during Injection Mold

Numerical simulation is used to solve the heat balance equation and evaluate a cost (or objective) function related to part quality and/or productivity. Several numerical methods such as Finite Element Method (FEM) [1] or Boundary Element Method (BEM) [2] can be used. Using BEM makes it possible to reduce by one the dimension of the mesh involved, meshing only the contour surface.

Mold-Cooling Optimization Approaches

The modeling of the heat transfer using BEM makes it possible to re-mesh only the surface of the channels during the optimization. An optimization method is used to modify the location and shape parameters of the channels and improve the cooling performance of the mold.

L. Silva [3] uses 3D FEM software to model injection molding cycle for complex geometries of molds. However, the computational burden of the 3D renders the integration of optimization difficult.

Mathey [4] used 2D stationary and transient BEM simulations of injection molds coupled with an SQP algorithm (Sequential Quadratic Programming) [5] to improve mold injection cooling.

In this paper, we extend Mathey's approach in view of addressing 3D optimization. An other

contribution is the improvement of the optimization formulation. We perform numerical simulations on a semi-industrial injection mold designed for the European project: *Eurotooling 21*. We optimize cooling channel locations on both 2D and 3D models. We attempt at minimizing the maximal cooling temperature of the plastic part subject to a temperature uniformity constraint.

STATIONARY HEAT TRANSFER PROBLEM

The variation of the temperature in the injection mold in production, after some cycle, is a quasistationary mode. After the average temperature of the mold during a cycle is stabilized (Fig. 1), we consider a stationary regime. We neglect the transitory oscillations of the temperature.



FIGURE 1. Schematic temperature evolution in injection mold during several cycles.

Solving Heat Balance Equation using Boundary Elements Method

The stationary heat conduction problem assuming constant thermal conductivity reduces to Laplace equation.

$$\Delta T = 0 \tag{1}$$

Multiplying this equation by a weighting function T*, and using Green's theorem, we obtain the well-known Somigliana's equation [6]:

$$C.T(P) + \int_{\Gamma} T.(\nabla T^*.n).d\Gamma = \int_{\Gamma} (\nabla T.n).T^*.d\Gamma \quad (2)$$

T is the temperature, and *C* is equal to 1 inside the domain and to 0.5 on its regular boundary. T^* and q^* denotes the fundamental solutions of the problem so-called Green's functions [6]:

$$2D \begin{cases} T^* = \frac{-1}{2\pi} \ln r & \\ q^* = \frac{-r.n}{2\pi r} & 3D \end{cases} \begin{cases} T^* = \frac{1}{2\pi r} \\ q^* = \frac{-r.n}{2\pi r^2} \end{cases}$$
(3)

After reorganization of equation (2), we obtain a non-symmetrical linear system.

Boundary Conditions

Figure 2 displays the boundary conditions on the mold.





The temperature of the coolant is T_C and the heat transfer coefficient, h, is related to the coolant flow rate (via Colburn correlation coefficient).

$$-\lambda \nabla T . n = h (T - T_C)$$
⁽⁴⁾

On the cavity surface, the flux density is calculated from the cycle time and polymer properties (heat due to polymer crystallization is neglected) [8].

Thermo Physical Parameters

All the parameters necessary for the numerical heat transfer simulations are referenced in Table 1.

TABLE 1. Thermo physical parameters for the mold and polymer.

| | Polymer (PP) | Mold (steel 40cmd8s) |
|---|--------------|-------------------------|
| $\lambda [W.m^{-1}.K^{-1}]$ | 0.63 | 34 |
| ρ [Kg.m ⁻³] | 891 | 7800 |
| Cp [J.Kg ⁻¹ .K ⁻¹] | 2740 | 460 |

| TABLE 2. Process parameters | | | | |
|-----------------------------|------------------|---------------------|--|--|
| T°C of injection | T°C of ejection | Time of injection = | | |
| $= 240^{\circ}\mathrm{C}$ | $= 100^{\circ}C$ | 20 s | | |

Mold Cooling Optimization

Figure (3) displays the coupling between the thermal solver and the optimization algorithm. Heat transfer computation is coupled with optimization method in order to modify automatically parameters. BEM simulation is performed and cost function is computed. The optimization method allows updating parameters with respect to constraints until a minimum of the cost function is reached. Matlab® SQP, based on the Newton method, is used for the optimization of continuous non-linear functions with continuous non-linear constraints.



FIGURE 3. Heat transfer simulation coupled with optimization algorithm.

Cost Function and Constraints

SQP algorithm minimizes just one objective. One method can consider the cost function as the weighting of two objectives, but this method imposes to choose a weighting coefficient. We propose to use the first objective as an optimization criterion (5), and the second as a non-linear constraints (eq 6):

$$\Phi = \max(T) \tag{5}$$

and
$$0 \le \sum_{i} (T_i - T_{moy})^2 \le 4$$
 (6)

The first criterion is the maximum of the temperature on the cavity surface (5).

The second criterion (6) improves the uniformity. For that, Park [9] proposes to minimize the variation of temperature distribution on the cavity surface compared to its average temperature. This function is defined as non-linear constraints. The final configuration provides an uniform and high average temperature, which give a very long cooling time.

Optimization Parameters

In this paper, we consider circular cooling channels in 2D and cylinder in 3D, 3 parameters are needed to describe each cooling channel. Our methodology can also account for optimizing non-linear geometries parameters. These processing parameters can define for instance the thermal regulation: temperature and flow rate of the coolant.

APPLICATION 1: RESULTS FOR A 2D CASE

In order to validate our approach, we consider the 2D injection geometry as described in Table 1 and 2.

Geometrical Parameters

Our methodology allows for various industrials constraints such as the definition of forbidden zones where one cannot put the cooling channels (due for the presence of the ejectors, or to keep the cooling channel within the mold, or to avoid inter-channel collisions). δ is the possible displacement according to axis X and Y for the channels.



FIGURE 4. Definition of the geometry parameters.

Numerical Parameters

| TABLE 3. | Numerical | parameters | for 2D | geometry. |
|----------|---------------|------------|--------|-----------|
| TTDDD U | 1 (uniterioui | parameters | 101 20 | Goomon . |

| Elements type | Linear |
|------------------------|---------|
| Number of elements | 1013 |
| Number of nodes | 1014 |
| Number of internal not | les 489 |

Optimization Parameters

Recall that the stationary BEM is used to evaluate the objective function at the mold cavity surface.

The optimization parameters for the locating of the ith cooling channel are (xi,yi) (see Figure 4). Thus, this 2D problem involves 16 optimization variables.

TABLE 4. Channels positions before and after optimization.

| | Initial values | | Final values | |
|-------------|----------------|-------|--------------|--------|
| Channel «i» | Χ | Y | X | Y |
| 1 | 0.135 | 0.018 | 0.129 | 0.014 |
| 2 | 0.11 | 0.018 | 0.106 | 0.014 |
| 3 | 0.067 | 0.018 | 0.063 | 0.014 |
| 4 | 0.05 | 0.018 | 0.046 | 0.0198 |
| 5 | 0.037 | 0.049 | 0.033 | 0.045 |
| 6 | 0.067 | 0.049 | 0.063 | 0.05 |
| 7 | 0.1 | 0.049 | 0.0984 | 0.0459 |
| 8 | 0.14 | 0.049 | 0.136 | 0.0496 |

TABLE 5. CPU timed for the direct 2D computing and the optimization.

| Meshing | Matrix A construction | Solving the system | Internals points computing |
|---------|-----------------------|--------------------|----------------------------------|
| 0.75 s | 1.2188 s | 0.64063 s | 11.0313 s |

The Table 5 displays the CPU time repartition for a direct computing. The matrix construction represents 45% of the direct computing.



FIGURE 5. Objective function versus iterations.

| TABLE 6. CPU | time for the 2 | D optimization | problem |
|--------------|----------------|----------------|-------------|
| Optimization | Iterations | Number of co | st function |

| time | Iterations | evaluations |
|--------|------------|-------------|
| 1320 s | 17 | 367 |
| | | |

As illustrated Figure 5, after only 3 optimizations iterations, the objective-function value is already within 5% from the final (optimized) value. The cost function is divided by the initial value to insure $0 \le \varphi \le 1$.



FIGURE 6. Temperature at the surface of the 2D mold cavity before and after optimization.

Figure 6 displays the temperature distribution along the mold cavity surface before and after optimization. We observe that both temperature variance and temperature average decreased significantly. Figure 7 shows the temperature distribution inside the mold temperature.



FIGURE 7. Temperature distribution after optimization for the 2D approach.

APPLICATION 2: SIMPLIFIED APPROACH 3D

We now report computational results on a 3D plastic part whose features are displayed on Figure 8 (unit in mm). The forbidden zones and the input data are identical to those of the 2D case.



FIGURE 8. Schema of plastic part.

Meshing Parameters

The data for the mesh are reported in Table 7.

| TABLE 7. Meshing parameters | for 3D geometry. |
|-----------------------------|------------------|
| Elements type | Linear |

| Number of elements | 3266 |
|--------------------|------|
| Number of nodes | 3204 |

The geometrical optimization parameters are now the coordinates of points P1 and P2 (see Figure 9).

The shape of the channels is assumed to be cylindrical and to have a rigid-body motion. Indeed, P2 can then be expressed in terms of P1 coordinates and of the constant channel length L (Figure 9):



FIGURE 9. Channel geometric parameters.

Thus, the optimization parameters for positions of the ith cooling channel are xi, yi, zi. Our 3D problem therefore involves 24 optimization variables.

| | Initial values | Init | ial valuas |
|---------------|------------------|------------|------------|
| optimization. | | | |
| TABLE 8. ch | annels positions | before and | after |

| Channel | Initial v | Initial values | | ues |
|---------|-----------|----------------|--------|--------|
| | Х | Y | Χ | Y |
| 1 | 0.135 | 0.018 | 0.131 | 0.021 |
| 2 | 0.11 | 0.018 | 0.101 | 0.02 |
| 3 | 0.067 | 0.018 | 0.058 | 0.02 |
| 4 | 0.05 | 0.018 | 0.046 | 0.019 |
| 5 | 0.037 | 0.049 | 0.033 | 0.045 |
| 6 | 0.067 | 0.049 | 0.063 | 0.05 |
| 7 | 0.1 | 0.049 | 0.0984 | 0.0459 |
| 8 | 0.14 | 0.049 | 0.136 | 0.0496 |

Table 9 displays the CPU time for a direct computational. The matrix construction represent 80% of the CPU time for the direct computing.

TABLE 9. CPU time for the direct 2D computing and the optimization.

| Meshing | Matrix A construction | Resolution of the system | CPU total |
|---------|-----------------------|--------------------------------|-----------|
| 0.75 s | 72.21 s | 13.295s | 86.145 s |
| | | | |

Table 10 displays the CPU time optimization.

| TABLE 10. | CPU 1 | time | for the | 2D o | ptim | ization |
|-----------|-------|------|---------|------|------|---------|
|-----------|-------|------|---------|------|------|---------|

| Optimization time | Iterations | Evaluation's number of the cost function |
|----------------------|------------|--|
| 13.5 H | 24 | 576 |

Sensibility of the Cost Function

As is illustrated Figure 10, the speed convergence of the cost function is fast.



FIGURE 10. Objective function versus iterations.

As it is illustrated in Figure 10, the form of the convergence of the cost function is fast. The value of the cost function is reduced by 90% in 4 iterations. SQP cannot respecting the constraints during optimization if the converge more quickly. This is an advantage in the industrials problems because an initial location of the channels, respecting all the

constraints, is something difficult to find. This advantage can be a problem if the user considers the accuracy of the convergence sufficient and will stop optimization,

Temperature Distribution

Figure 11 displays the temperature gradient of the surface of mold cavity. The results are similar to 2D geometry, except for the boundary effect.



FIGURE 11. Surface temperature distribution at the surface of the mold cavity.

CONCLUSION

The BEM has been used to solve heat transfer equation during the cooling step of the molding process, for a 2D and 3D problem. This model is coupled with SQP optimization algorithm to find the best geometrical parameters according to a cost function. SQP is a mono-objective algorithm but it is possible to define one of the objectives like non-linear constraints. It is crucial to study the parameterization of the channels to reduce optimization variable and thus CPU time. The use of the elliptical cooling channels should be further investigated for more complex shapes in future works.

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NOMENCLATURE

M, P and C indexes refer respectively to the mold, the polymer part and the coolant.

- C: Heat capacity [J.Kg⁻¹.K⁻¹]
- d: Cooling channel diameter [m]
- h: Heat transfer coefficient [W.m⁻².K⁻¹]
- S: Surface [m²]
- q: Temperature gradient [K.m⁻¹]
- R: Thermal contact resistance [K.m².W⁻¹]
- Tc : Temperature of the coolant [s]
- T: Temperature [°C]
- V: Volume [m³]
- λ : Thermal conductivity [W.m⁻¹.K⁻¹]
- Φ : Heat flux [W.m²]
- ρ: Mass density [Kg.m⁻³]

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