Projects

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In case of errors, typos please inform me.

1 Introduction

I present two problems modeling the same situation - stationary i.e. modeling the situation after the temperature stabilizes and evolutionary - time dependent.

The easiest is 1D stationery with finite difference discretization. And this one is really recommended unless somebody is more ambitious or interested in doing more. Naturally, if one picks more complicated version I will be more lenient in the evaluating process - but anyway, any, not finished project will be not given a high grade (or many points) so any version you choose try to make it work.

2 Optimal control problem

A room $(0, a_1) \times (0, a_2)$ with the height H > 0 is heated by setting heating on a part of the wall. We want to know how to control the temperature in a part of the room by changing the temperature of the heater. The temperature is controlled by a given PDE. We have to set the boundary values such that the temperature in the room is as close as possible to a given function.

We have two cases - stationary and time-dependent. The simplest is a 1D version where we assume that we heat one wall and control the temperature along with the room (so the temperature equals u(x) where x is the distance from the wall).

We can use different discretizations for given models: I propose to use standard finite difference method (FDM), the domain is an interval in 1D or a rectangle in 2D. So we can use equidistant mesh in each direction.

2.1 Stationary case

Let $\Omega = \prod_{k=1}^{d} (0, a_k)^d d = 1, 2$, we heat $\Gamma_1 = \{0\}$ in 1D or

$$\Gamma_1 = \{(0,s) : 0 \le s \le a_2\},\$$

we assume that u satisfies:

$$-Lu - S(u) = -\sum_{k=1}^{d} \frac{\partial}{\partial x_k} A_k(t, x) \frac{\partial}{\partial x_k} u(x) - S(u) = f(x) \qquad x \in \Omega$$
(1)

$$\lambda \frac{\partial}{\partial n} u(s) + u(s) = g(s) \qquad x \in \Gamma_1 \tag{2}$$

$$\frac{\partial}{\partial n}u(s) = 0 \quad x \in \Gamma_2 = \partial\Omega \setminus \Gamma_1 \tag{3}$$

Let simplify the model and set $A_k(t, x) = \alpha > 0$ a positive constant, λ is also a positive constant, S(u) a given function (we should be able to use any univariate function. u - the solution depends on the function g defined over Γ_1 , we assume $g_{min} \leq g(s) \leq g_{max}$. We want to compute g_0 such that

$$F(g_0) = \min_g F(g) \quad F(g) = \int_D |u(x) - H(x)|^2 dx$$

where H is a given function defined over $D = \prod_{k=1}^{d} [b_k, a_k]$, here $0 < b_k < a_k$.

Summing up - we define the function $F(g) = \int_D |u(x) - H(x)|^2 dx$ with u solving the BVP and we would like to find g_0 minimizing this function

Possible discretizations:

- FDM the derivative in the Robin boundary condition may be approximated by the simplest 2 points finite difference (forward or backward) that's the simplest and recommended
- linear continuous finite element method (FEM)
- quadratic continuous finite element method (FEM) (only if you are really interested in those methods)
- another method e.g. spectral collocation method or a finite volume method (only if you are really interested in those methods)

2.2 Non-stationary case

We heat the same part of the boundary, cf. Section 2.1 and that u(t, x) satisfies: equation:

$$\frac{\partial}{\partial t}u - \sum_{k=1}\frac{\partial}{\partial x_k}A_k(t,x)\frac{\partial}{\partial x_k}u(t,x) - S(u) = 0 \qquad x \in \Omega \quad t \in (0, T_{Max})$$
(4)

$$\lambda \frac{\partial}{\partial n} u(t,s) + u(t,s) = g(t,s) \qquad x \in \Gamma_1 \quad t \in (0, T_{Max})$$
(5)

$$\frac{\partial}{\partial n}u(t,s) = 0 \quad x \in \Gamma_2 = \partial\Omega \setminus \Gamma_1 \quad t \in (0, T_{Max})$$
(6)

 S, a_k are the same as in Section 2.1

u is controlled by $g^*(t, x)$ defined for $x \in D$ s.t. $g_{min} \leq g^* \leq g_{max}$ and

$$F(g^*) = \min_{g} F(g) \quad F(g) = \int_0^{T_{max}} \int_D |u(t,x) - H(t,x)|^2 dx \, dt$$

where H is a given function defined on $(0, T_{max}) \times D$ for $D = \prod_{k=1}^{d} [b_k, a_k], 0 < b_k < a_k d = 1, 2$. We will further simplify the model i.e. we take

$$g(t,(0,s)) = \begin{cases} g_0(t) & s \in [0, 0.25 * a_2] \\ g_0(t)(1 - 0.5 \frac{s - 0.25 * a_2}{0.75 * a_2}) & s \in [0.25 * a_2, a_2] \end{cases}$$
(7)

Thus really we are looking for $g_0: (0, T_{Max}) \to R$ which minimizes F.

We introduce a space discretization (FDM or FEM) obtaining a IVP which we solve

- using any octave solver e.g. **lsode**() (default)
- using implicit Euler or Crank-Nicholson scheme

In case of using lsode() we have to define $g_0(t)$ for any time $t \leq T_{Max}$ even if we set discrete times since the octave solver may need to compute the vector field also at auxiliary times. The most reasonable is to take g_0 as a linear spline defined by the values at our discrete times (e.g. equidistant points on $[0, T_{MAX}]$. So our unknown will be the values of g_0 at discrete times.

2.2.1 Tests

Please test your code for $a_1 = 2, a_2 = 4, b_1 = 1, b_2 = 2, A_k = \alpha \equiv 1, S(u) \equiv 0$ with f = 2 or (this is optional if you are interested in it only) $S(u) = -atan(u)/\pi$ with $f = 0, \lambda = 1, H \equiv 20$. In 1D case take d = 1 $a_1 = 2, b_1 = 1$.

One can show me the code - how it works and write a 2-3 pages long report, describing the discretization and used tools (e.g. what optimization or ODE solver was used, etc). You should also test parts of the solver e.g. if the solver for the respective PDE works for some known solutions etc. I can ask that during the exam.