

1 Optimal control with ODE constraints

Lets try to derive the adjoint equation and the gradient, when we let $E(y, u)$ be the following ODE:

$$\begin{cases} y'(t) = \alpha y(t) + u(t), & t \in (0, T) \\ y(0) = y_0 \end{cases}$$

We also choose the functional to be

$$J(y, u) = \frac{1}{2} \int_0^T u(t)^2 dt + \frac{1}{2} (y(T) - y^T)^2$$

Before we calculate the different terms in the gradient, we want to fit our ODE into an expression E . we do this by "moving" the initial condition into the equation:

$$E(y, u) = y' - \alpha y - u + \delta_0(y - y_0)$$

Here the δ_0 means evaluation at 0. Now lets find E_u , E_y , $\langle J_u(u), s \rangle$ and J_y with respect to our E and J .

$$E_u(y(u), u) = -1$$

$$E_y(y(u), u) = \frac{\partial}{\partial t} - \alpha + \delta_0, \text{ where } \delta_0 \text{ is evaluation at } 0$$

$$\langle J_u(u), s \rangle = \int_0^T u(t)s(t)dt$$

Lets be more thorough with J_y , which is the right hand side in the adjoint equation.

$$\begin{aligned} J_y(y(u), u) &= \frac{\partial}{\partial y} \left(\frac{1}{2} \int_0^T u^2 dt + \frac{1}{2} (y(T) - y^T)^2 \right) \\ &= \frac{\partial}{\partial y} \frac{1}{2} (y(T) - y^T)^2 \\ &= \frac{\partial}{\partial y} \frac{1}{2} \left(\int_0^T \delta_T (y - y^T) dt \right)^2 \\ &= \delta_T \int_0^T \delta_T (y(t) - y^T) dt \\ &= \delta_T (y(T) - y^T) = L \end{aligned}$$

We have $E_y(y(u), u) = \frac{\partial}{\partial t} - \alpha + \delta_0$, but for the adjoint equation we need to find E_y^* . To derive the adjoint of E_y , we will apply it to a function v and then take the L^2 inner product with another function w . The next step is

then to try to "move" the operator E_y from v to w . As becomes clear below, partial integration is the main trick to achieve this:

$$\begin{aligned}
\langle E_y v, w \rangle &= \int_0^T (v'(t) - \alpha v(t) + \delta_0 v(t)) w(t) dt \\
&= \int_0^T v'(t) w(t) dt - \alpha \int_0^T v(t) w(t) dt + v(0) w(0) \\
&= - \int_0^T v(t) w'(t) dt + v(t) w(t) \Big|_0^T - \alpha \langle v, w \rangle + v(0) w(0) \\
&= - \int_0^T v(t) w'(t) dt - \alpha \langle v, w \rangle + v(T) w(T) \\
&= \langle v, Pw \rangle
\end{aligned}$$

Where $P = -\frac{\partial}{\partial t} - \alpha + \delta_T$. This means that $E_y^* = P$, and we now have the left hand side in the adjoint equation. The right hand side is $J_y(y(u), u) = L$, which we have already found. If we write the adjoint equation on variational form it will look like this: $\langle Pp, w \rangle = \langle L, w \rangle$. To get back to standard ODE form, we can do some manipulation:

$$\begin{aligned}
\langle -p' - \alpha p + \delta_T p, w \rangle &= \langle \delta_T(y(T) - y^T), w \rangle \\
\langle -p' - \alpha p, w \rangle &= \langle \delta_T(y(T) - y^T - p), w \rangle
\end{aligned}$$

The right hand side is point evaluation at $t = T$, while the left hand side is an expression for all t . This finally gives us our adjoint equation:

$$\begin{cases} -p'(t) = \alpha p(t) \\ p(T) = y(T) - y^T \end{cases}$$

This is a simple and easily solvable ODE.

Expression for the gradient

We now have all the ingredients for finding an expression for the gradient of \hat{J} . If we remember that $\langle \hat{J}'(u), s \rangle = \langle y'(u)^* J_y(u), s \rangle + \langle J_u(u), s \rangle$, and all the different expressions for all the terms we calculated, we find:

$$\begin{aligned}
\langle \hat{J}'(u), s \rangle &= \langle y'(u)^* J_y(u), s \rangle + \langle J_u(u), s \rangle \\
&= \langle -E_u^* p, s \rangle + \langle J_u(u), s \rangle \\
&= \langle -(-1)^* p, s \rangle + \langle u, s \rangle \\
&= \langle p + u, s \rangle \\
&= \int_0^T (p(t) + u(t)) s(t) dt
\end{aligned}$$

Note that the adjoint of a constant is just the constant itself.