

# 1 Periodic Boundary Conditions

We consider the advection diffusion equation with periodic boundary conditions and a corresponding OCP:

$$\begin{aligned} & \min \frac{1}{2} \|\rho - \hat{\rho}\|^2 + \frac{\beta}{2} \|\mathbf{w}\|^2 \\ & \text{subject to:} \\ & \frac{\partial \rho}{\partial t} = \frac{\partial^2 \rho}{\partial x^2} - \frac{\partial \rho \mathbf{w}}{\partial x} \\ & \rho(a) = \rho(b) \\ & \frac{\partial \rho(a)}{\partial x} - \rho(a) \mathbf{w}(a) = -\frac{\partial \rho(b)}{\partial x} + \rho(b) \mathbf{w}(b) \end{aligned}$$

The relevant part of the Lagrangian is then:

$$\begin{aligned} \mathcal{L} = & \dots - \int_0^T \int_{\Omega} \left( \frac{\partial \rho}{\partial t} - \frac{\partial^2 \rho}{\partial x^2} + \frac{\partial \rho \mathbf{w}}{\partial x} \right) q dr dt \\ & - \int_0^T \left( \rho(b) q_1 - \rho(a) q_1 + \frac{\partial \rho(b)}{\partial x} q_2 - \rho(b) \mathbf{w}(b) q_2 + \frac{\partial \rho(a)}{\partial x} q_2 - \rho(a) \mathbf{w}(a) q_2 \right) dt. \end{aligned}$$

Taking partial derivatives, the relevant part of the Lagrangian is:

$$\mathcal{L} = \dots - \int_0^T \left[ q \frac{\partial \rho}{\partial x} - \rho \frac{\partial q}{\partial x} - \rho \mathbf{w} q \right]_a^b - \left( \rho(b) q_1 - \rho(a) q_1 + \frac{\partial \rho(b)}{\partial x} q_2 - \rho(b) \mathbf{w}(b) q_2 + \frac{\partial \rho(a)}{\partial x} q_2 - \rho(a) \mathbf{w}(a) q_2 \right) dt.$$

Taking the derivative with respect to  $\rho$  gives:

$$\begin{aligned} \mathcal{L}_{\rho} h = & \dots - \int_0^T \left[ q \frac{\partial h}{\partial x} - h \frac{\partial q}{\partial x} - h \mathbf{w} q \right]_a^b \\ & - \left( h(b) q_1 - h(a) q_1 + \frac{\partial h(b)}{\partial x} q_2 - h(b) \mathbf{w}(b) q_2 + \frac{\partial h(a)}{\partial x} q_2 - h(a) \mathbf{w}(a) q_2 \right) dt \end{aligned}$$

Writing all terms explicitly:

$$\begin{aligned} \mathcal{L}_{\rho} h = & \dots + \int_0^T \left( -q(b) \frac{\partial h(b)}{\partial x} + h(b) \frac{\partial q}{\partial x} + h(b) \mathbf{w}(b) q(b) + q(a) \frac{\partial h(a)}{\partial x} - h(a) \frac{\partial q(a)}{\partial x} - h(a) \mathbf{w}(a) q(a) \right. \\ & \left. - h(b) q_1 + h(a) q_1 - \frac{\partial h(b)}{\partial x} q_2 + h(b) \mathbf{w}(b) q_2 - \frac{\partial h(a)}{\partial x} q_2 + h(a) \mathbf{w}(a) q_2 \right) dt \end{aligned}$$

Then considering the terms that satisfy  $\frac{\partial h}{\partial x} \neq 0$  we get:

$$-q(b) + q(a) - q_2(b) - q_2(a) = 0,$$

so that  $q(b) = -q_2(b)$  and  $q(a) = q_2(a)$ . For  $h \neq 0$  we have:

$$\frac{\partial q(b)}{\partial x} + \mathbf{w}(b) q(b) - \frac{\partial q(a)}{\partial x} - \mathbf{w}(a) q(a) - q_1(b) + q_1(a) + \mathbf{w}(b) q_2(b) + \mathbf{w}(a) q_2(a) = 0$$

Using the first condition, we get:

$$\frac{\partial q(b)}{\partial x} + \mathbf{w}(b)q(b) - \frac{\partial q(a)}{\partial x} - \mathbf{w}(a)q(a) - q_1(b) + q_1(a) - \mathbf{w}(b)q(b) + \mathbf{w}(a)q(a) = 0,$$

so that terms involving  $\mathbf{w}$  cancel to give:

$$\frac{\partial q(b)}{\partial x} - \frac{\partial q(a)}{\partial x} - q_1(b) + q_1(a) = 0.$$

If  $q_1 = q_2$  we get:

$$\frac{\partial q(b)}{\partial x} - \frac{\partial q(a)}{\partial x} + q(b) + q(a) = 0.$$

## 2 Time independent control

We wanted to see whether the time independent flow control is similar to the  $\nabla V_{ext}$  of the target. The target state was influenced by  $V_{ext} = 0.1y_2$ . The forward state for the OCP was influenced by  $V_{ext} = 0.01y_2$ . Figure 1 shows the control and  $\nabla V_{ext}$  of the target. Why is one positive and one negative? Is it because they are opposite signs in PDE?

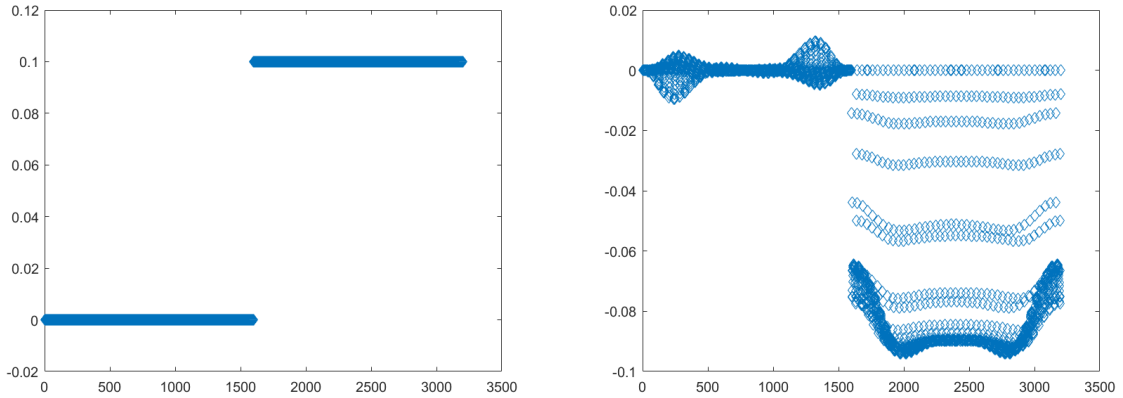


Figure 1:  $\nabla V_{ext}$  of target and optimal control

## 3 Multishape Channel

The target flow profile works now and can be seen in Figure 2. We chose  $N = 20$ ,  $n = 20$  for each shape and  $T = 5$ . The optimal control problem is still not quite working. For  $\beta = 10^{-1}$ , it seems to work or almost, but for  $10^{-3}$  it converges in three iterations, where the last error is zero. However,  $J_{FW} < J_{Opt}$ . When decreasing  $\lambda$  from 0.01 to 0.001, the convergence steps are smaller and hopefully therefore it will converge to a minimum. The target and optimal  $\rho$  for  $\beta = 10^{-1}$  are displayed in Figure 3. We have  $J_{FW} = 0.2092$  and  $J_{Opt} = 0.1563$ .

- exact problem not exact yet - multishape JFW ; JOpt

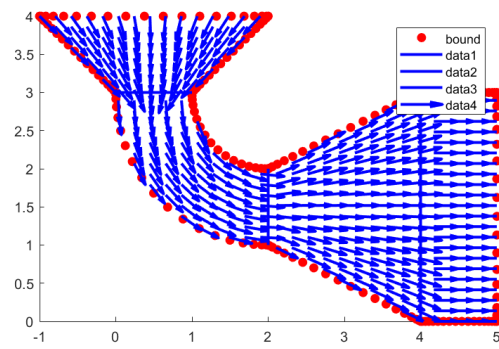


Figure 2: Target flow setup

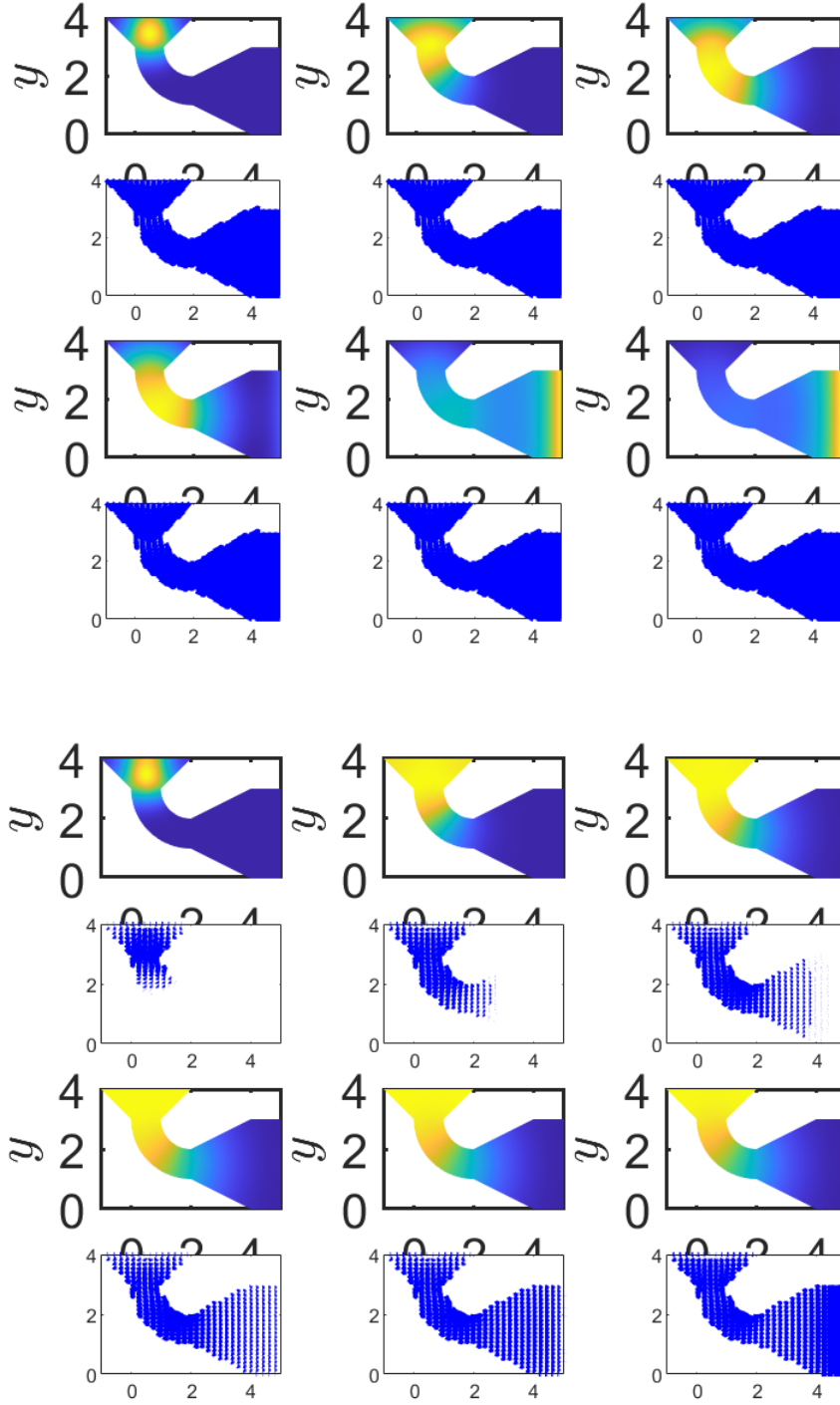


Figure 3:  $\hat{\rho}$  and optimal  $\rho$ ,  $\beta = 10^{-1}$ .