

# PARAMETER ESTIMATION WITH NEURAL NETWORKS SUBJECT TO PDE CONSTRAINTS

Consider

$$(1) \quad \nabla \cdot (k(x) \nabla u(x)) = 0$$

subject to the appropriate BCs.

Here,  $k(x)$  is the unknown coefficient. We assume that  $N_k$  measurements of  $k$  and  $N_u$  measurements of  $u$  are available:  $k^*(x_i)$  ( $i = 1, \dots, N_k$ ) and  $u^*(x_i)$  ( $i = 1, \dots, N_u$ ).

Define NNs for  $k(x)$  and  $u(x)$ ,  $\hat{k}(x; \gamma) = \mathcal{NN}_k(x; \gamma)$  and  $\hat{u}(x; \theta) = \mathcal{NN}_u(x; \theta)$ . Substituting this in the government equation yields:

$$(2) \quad f(x; \gamma, \theta) = \nabla \cdot (\hat{k}(x) \nabla \hat{u}(x)) = \mathcal{NN}(x; \gamma, \theta).$$

Then, we define the loss function as:

$$(3) \quad \mathcal{L}(\theta, \gamma) = \frac{1}{N_k} \sum_{i=1}^{N_k} (\hat{k}(x_i; \gamma) - k^*(x_i))^2 + \frac{1}{N_u} \sum_{i=1}^{N_u} (\hat{u}(x_i; \theta) - u^*(x_i))^2 + \frac{1}{N_c} \sum_{i=1}^{N_c} f(x_i; \gamma, \theta)^2$$

In the last term, the  $N_c$  collocation points could be chosen uniformly or non-uniformly depending on the problem.