

Let  $J(V, W)$  be any functional on  $V \times W$  with  $V$  and  $W$  be two finite dimensional spaces. Say we have some assumptions on  $J$ ,  $V$  and  $W$  to ensure the required minimums can be reached. Then we have

$$\min_{v \in V} \min_{w \in W} J(v, w) = \min_{(v, w) \in V \times W} J(v, w).$$

**Proof.** Let  $(v_0, w_0)$  and  $(v_1, w_1)$  be two points in  $V \times W$  such that

$$J(v_1, w_1) = \min_{v \in V} \min_{w \in W} J(v, w), \quad J(v_0, w_0) = \min_{(v, w) \in V \times W} J(v, w).$$

Then obviously,  $J(v_0, w_0) \leq J(v_1, w_1)$ . We now prove that  $J(v_1, w_1) \leq J(v_0, w_0)$ .

Let  $F(v) = \min_{w \in W} J(v, w)$ , then

$$J(v_1, w_1) = \min_{v \in V} F(v) = F(v_1) \leq F(v_0) = \min_{w \in W} J(v_0, w) \leq J(v_0, w_0),$$

that is the required result.

First note: please replace  $\gamma_{\partial F}$  in the definition of the wirebasket coarse interpolant by  $\gamma_{h, \partial F}$ . All the proofs are still valid by Lemma 4.20.

To solve (8.9), we first express any  $w_0 \in V_0(\Gamma)$  into

$$w_0 = \sum_F \gamma_{h, \partial F}(w_0) \phi_F + \sum_E \sum_{x_j \in E} w_0(x_j) \phi_j + \sum_k w_0(x_{v_k}) \phi_k,$$

where  $\phi_F = I_F^0 1$  and  $\phi_j$  for  $x_j \in \mathcal{W}_i$  are natural nodal basis functions on  $\mathcal{W}_i$ . We see

$$\langle g_0, w_0 \rangle = \sum_F \gamma_{h, \partial F}(w_0) \langle g_0, \phi_F \rangle + \sum_E \sum_{x_j \in E} w_0(x_j) \langle g_0, \phi_j \rangle + \sum_k w_0(x_{v_k}) \langle g_0, \phi_k \rangle,$$

using  $\gamma_{h, \partial F}(w_0) = \frac{1}{n(\partial F)} \sum_{x_j \in \partial F} w_0(x_j)$ ,  $n(\partial F)$  is the number of nodes on  $\partial F$ , we can write

$$\langle g_0, w_0 \rangle = x^\top b$$

with  $x$  a vector consisting of all wirebasket nodal values  $w_0(x_j)$ . Let  $x_i$  be a vector with all components of  $x$  corresponding to  $\mathcal{W}_i$ , and  $e_i$  a vector with the same dimension as  $x_i$  but with all components 1. Then using the definition of  $\|\cdot\|_{h, \mathcal{W}_i}$ , (8.9) can be rewritten as follows

$$\min_x \frac{1}{2} \sum_i \rho_i h \min_{\lambda_i \in R^1} (x_i - \lambda_i e_i)^\top (x_i - \lambda_i e_i) - x^\top b,$$

which is equivalent to the following linear systems obtained by taking the derivatives with respect to  $x$  and  $\lambda_i$ :

$$(1) \quad \sum_i \rho_i h (x_i - \lambda_i e_i) = b,$$

$$(2) \quad e_i^\top \rho_i h (x_i - \lambda_i e_i) = 0, \quad i = 1, \dots, p.$$

Let  $B$  be the matrix such that  $Bx = \sum_i \rho_i h x_i$ , then eliminating  $x$  from the equation (1), we derive the equation with respect to  $\lambda_i$ :

$$(e_i^\top e_i) \lambda_i - e_i^\top B^{-1} \sum_j (\rho_j h e_j) \lambda_j = e_i^\top B^{-1} b, \quad i = 1, \dots, p.$$

Once we get  $\lambda_i$  from the above equation, we can then solve (1) to get  $x$ .