

# The Best Interpolation Problem

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## 1 The Best Interpolation Problem

Consider the problem of minimizing the functional

$$I(\mathbf{u}) = \int_0^1 u^2(t) dt \quad (1.1)$$

over the admissible set

$$A = \{\mathbf{u} \in L^2[0, 1] : J(\mathbf{u}) = C\}, \quad (1.2)$$

where the vector valued linear function  $J : L^2[0, 1] \rightarrow \mathbb{R}^n$  is defined to be

$$J(\mathbf{u})_k = J_k(\mathbf{u}) = \int_0^1 \psi_k(t)u(t) dt \quad (1.3)$$

for given functions  $\psi_k \in L^2[0, 1]$ . The vector  $C \in \mathbb{R}^n$  belongs to the range of  $J$ , i.e.  $C_i = \int_0^1 \psi_i(t)\hat{x}(t) dt$  for some  $\hat{x} \in L^2[0, 1]$ .

Note that the admissible set  $A$  is convex, closed and nonempty. The existence of minimum of the functional  $I$  over the set  $A$  follows from the existence of the closest point to the origin in the closed convex set  $A \subset L^2[0, 1]$ .

References for such problems are in books such as: [6, 3, 1, 4, 5]

## 2 A Necessary Condition for the Minimizer. The Exact Solution

We will derive a necessary condition for a function  $\mathbf{u} \in A$  to be a minimum.

**Theorem 2.1 (Lagrange Multipliers Rule)** *Suppose that the set of constraints  $J(\mathbf{u}) = C$  in minimization problem*

$$\min I(\mathbf{u}) \quad \text{subject to } \mathbf{u} \in A \quad (2.1)$$

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satisfies the following property. There exists a set of  $L^2[0, 1]$  functions  $w_1, \dots, w_n$  such that the matrix  $Y \in \mathbb{R}^{n \times n}$  with the components

$$Y_{ij} = J_i(w_j) = \int_0^1 \psi_i(t)w_j(t) dt \quad (2.2)$$

is nonsingular.

Suppose that  $u \in A$  is a solution of problem (2.1). Then there exists a vector  $\lambda \in \mathbb{R}^n$  such that for every  $v \in L^2[0, 1]$

$$\int_0^1 u(t)v(t) dt = \sum_{j=1}^n \lambda_j \int_0^1 \psi_j(t)v(t) dt. \quad (2.3)$$

**Proof.** This proof uses the idea from the proof of theorem 2 from chapter 8.4 of the book "Partial Differential equations" by L. C. Evans [2].

Let  $v \in L^2[0, 1]$ . Consider the function  $j : \mathbb{R}^{n+1} \rightarrow \mathbb{R}^n$  with components

$$j_k(\tau, \sigma_1, \dots, \sigma_n) = J_k(\tau v + \sigma_1 w_1 + \dots + \sigma_n w_n) \quad (2.4)$$

$$= \int_0^1 \psi_k(t)(\tau v(t) + \sigma_1 w_1(t) + \dots + \sigma_n w_n(t)) dt, \quad (2.5)$$

where  $k = 1, \dots, n$  and the functions  $w_1, \dots, w_n \in L^2[0, 1]$  satisfy the condition from the hypothesis. Note that  $j(0, \dots, 0) = 0 \in \mathbb{R}^n$  and  $D_{\sigma} j(0, \dots, 0) = Y$  is nonsingular. Therefore, by the Implicit Function Theorem, there exists a differentiable function  $\phi : \mathbb{R} \rightarrow \mathbb{R}^n$  defined in a neighborhood  $Q = \{\tau \in \mathbb{R} : |\tau| \leq \tau_0\}$  of the origin such that

$$j(\tau, \phi(\tau)) = 0 \quad (2.6)$$

for all  $\tau \in Q$ . Moreover,

$$\phi'(0) = -[D_{\sigma} j(0, \dots, 0)]^{-1} D_{\tau} j(0, \dots, 0) = -Y^{-1}b, \quad (2.7)$$

where  $b \in \mathbb{R}^n$  is the vector with the components  $b_k = \int_0^1 \psi_k(t)v(t) dt$ . Now, consider the variation  $u + \tau v + \phi(\tau) \cdot w$  of the function  $u$ . Because  $J(u + \tau v + \phi(\tau) \cdot w) = J(u) + j(\tau, \phi(\tau)) = C$ , this variation belongs to the admissible set  $A$ . Therefore, the function  $i(\tau) := I(u + \tau v + \phi(\tau) \cdot w)$  attains its local minimum at the origin, which implies the equality  $i'(0) = 0$ . The latter equality can be recast in the form

$$\int_0^1 u(t)(v(t) + \phi'(0) \cdot w(t)) dt = 0.$$

Hence,

$$\int_0^1 u(t)v(t) dt = - \int_0^1 \phi'(0) \cdot w(t) dt. \quad (2.8)$$

Using expression (2.7) for  $\phi'(0)$ , we find that

$$\phi'(0) \cdot w = -Y^{-1}b \cdot w = - \sum_{j=1}^n \sum_{i=1}^n [Y^{-1}]_{ij} \int_0^1 \psi_j(s)v(s) ds w_i. \quad (2.9)$$

Substituting this expression into equation (2.8), we obtain

$$\begin{aligned}\int_0^1 \mathbf{u}(t)\mathbf{v}(t) dt &= \int_0^1 \sum_{j=1}^n \sum_{i=1}^n [Y^{-1}]_{ij} \int_0^1 \psi_j(s)\mathbf{v}(s)w_i(t) ds dt \\ &= \sum_{j=1}^n \int_0^1 \sum_{i=1}^n [Y^{-1}]_{ij} w_i(t) dt \int_0^1 \psi_j(s)\mathbf{v}(s) ds.\end{aligned}\quad (2.10)$$

Define  $\lambda_j = \int_0^1 \sum_{i=1}^n [Y^{-1}]_{ij} w_i(t) dt$ . Then equation (2.10) can be rewritten in the form

$$\int_0^1 \mathbf{u}(t)\mathbf{v}(t) dt = \sum_{j=1}^n \lambda_j \int_0^1 \psi_j(t)\mathbf{v}(t) dt \quad (2.11)$$

as required. ■

**Remark 2.2** *The latter theorem holds if we assume that the functions  $\psi_i$ ,  $i = 1, \dots, k$ , are such that their Gram matrix  $Y = \{\langle \psi_i, \psi_j \rangle_{L^2[0,1]}\}_{i,j=1}^n$  is nonsingular.*

**Corollary 2.3** *Suppose that the Gram matrix  $Y$  of the constraint functions  $\psi_i$  given by  $Y = \{\langle \psi_i, \psi_j \rangle_{L^2[0,1]}\}_{i,j=1}^n$  is nonsingular. Then there exists a unique minimizer of the functional  $I$  in the admissible set  $A$ .*

*Moreover, the constant vector  $\lambda = (\lambda_1, \dots, \lambda_n)^T$  satisfying relation (2.3) from the previous theorem is given by the formula*

$$\lambda = Y^{-1}C,$$

where  $C = (C_1, \dots, C_n)^T$ .

*The minimum of the functional  $I$  over the admissible set  $A$  is given by the formula*

$$I_{\min} = \min_{\mathbf{u} \in A} I(\mathbf{u}) = C^T Y^{-1} C$$

*and is attained on the function  $\mathbf{u} = \sum_{i=1}^n \lambda_i \psi_i$ , where  $\lambda = Y^{-1}C$ .*

**Proof.** As it was already mentioned, the existence of a minimizer follows from the fact that the admissible set  $A$  is a nonempty closed convex set: there exists a closest point to the origin in  $A$ .

Let us prove the uniqueness. Suppose that  $\mathbf{u}_1, \mathbf{u}_2 \in A$  minimize the functional  $I$  over  $A$ . Then  $I(\mathbf{u}_1) = I(\mathbf{u}_2) = I_{\min}$ . Let us apply theorem 2.1 to the function  $\mathbf{u}_1$ . There exists  $\lambda \in \mathbb{R}^n$  such that for all  $\mathbf{v} \in L^2[0, 1]$

$$\int_0^1 \mathbf{u}_1(t)\mathbf{v}(t) dt = \sum_{i=1}^n \lambda_i \int_0^1 \psi_i(t)\mathbf{v}(t) dt. \quad (2.12)$$

Set  $\mathbf{v} = \psi_k$ , then

$$\int_0^1 \mathbf{u}_1(t)\psi_k(t) dt = \sum_{i=1}^n \lambda_i \int_0^1 \psi_i(t)\psi_k(t) dt$$

for  $k = 1, \dots, n$  or, equivalently,

$$Y\lambda = C,$$

which proves that  $\lambda = Y^{-1}C$ .

By repeating the same argument for  $u_2$ , we obtain that

$$\int_0^1 u_2(t)v(t) dt = \sum_{i=1}^n \lambda_i \int_0^1 \psi_i(t)v(t) dt \quad (2.13)$$

for all  $v \in L^2[0, 1]$ , where  $\lambda = Y^{-1}C$ . Comparing equations (2.12) and (2.13), we conclude that  $\int_0^1 u_1(t)v(t) dt = \int_0^1 u_2(t)v(t) dt$  for all  $v \in L^2[0, 1]$ , which implies that  $u_1 = u_2$  as functions in  $L^2[0, 1]$  (almost everywhere). By setting  $v = u_1$  in (2.12), we obtain

$$I_{\min} = I(u_1) = \int_0^1 u_1^2(t) dt = \sum_{i=1}^n \lambda_i C_i = C^T Y^{-1} C.$$

It remains to check that  $u = \sum_{i=1}^n \lambda_i \psi_i$  is the unique minimizer of  $I$  in the admissible set  $A$ . It is easy to see that  $u \in A$ . Indeed,

$$\int_0^1 \psi_k(t)u(t) dt = \sum_{i=1}^n \lambda_i Y_{ik} = C_k.$$

The value of the functional  $I$  on  $u$  is

$$I(u) = \int_0^1 \left( \sum_{i=1}^n \lambda_i \psi_i(t) \right)^2 dt \quad (2.14)$$

$$= \sum_{i=1}^n \sum_{j=1}^n Y_{ij} \lambda_i \lambda_j. \quad (2.15)$$

Substituting  $\lambda = Y^{-1}C$  into the latter equality, it is easy to obtain that

$$I(u) = C^T Y^{-1} C = I_{\min}.$$

The proof is complete. ■

**Example 2.4** *The Exact Solution for the Case of One Constraint.*

Consider the problem of minimizing the functional  $I(u) = \int_0^1 u^2(t) dt$  over the admissible set  $A = \{u \in L^2[0, 1] : \int_0^1 \psi(t)u(t) dt = C\}$ , where  $0 \neq \psi \in L^2[0, 1]$  and  $C = \int_0^1 \psi(t)\hat{x}(t) dt$  for some  $\hat{x} \in L^2[0, 1]$ . Without loss of generality, assume that  $C \geq 0$ . By corollary 2.3, the following lemma holds.

**Lemma 2.5** Consider the optimization problem (2.1), where the admissible set

$$A = \{u \in L^2[0, 1] : \int_0^1 \psi(t)u(t) dt = C\}$$

with  $0 \neq \psi \in L^2[0, 1]$  and  $C = \int_0^1 \psi(t)\hat{x}(t) dt \geq 0$ . Then the function  $u = \frac{C}{\|\psi\|_2^2}\psi$  minimizes the functional  $I$  over the admissible set  $A$ .

It is easy to prove the lemma without using corollary 2.3 or theorem 2.1. Indeed, for  $u = \frac{C}{\|\psi\|_2^2}\psi$  we have that  $I(u) = I(\frac{C}{\|\psi\|_2^2}\psi) = \frac{C^2}{\|\psi\|_2^2}$ . On the other hand, for every  $w \in A$  the following inequality holds:

$$C = \int_0^1 \psi(t)w(t) dt \leq \left( \int_0^1 w^2(t) dt \right)^{1/2} \left( \int_0^1 \psi^2(t) dt \right)^{1/2}. \quad (2.16)$$

Recall that  $C \geq 0$  (we have assumed it in the beginning; we can always recast the problem to an equivalent one with  $C \geq 0$ ). Hence, squaring inequality (2.16), we obtain

$$I(u) = \frac{C^2}{\int_0^1 \psi^2(t) dt} \leq \int_0^1 w^2(t) dt = I(w)$$

for all  $w \in A$ , which proves that  $u = \frac{C}{\|\psi\|_2^2}\psi$  is the global minimum of  $I$  over the admissible set  $A$ .

## References

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