

The Best Interpolation Problem

Oksana Bihun*

July 17, 2007

Continuous Optimization Workshop, MSRI
Notes on Project 4

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*

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

*University of Missouri-Columbia

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 ψ_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 ψ_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

- [1] F. Deutsch. *Best Approximation in Inner Product Spaces*. Springer-Verlag, New York, 2001.
- [2] L. Evans. *Partial Differential Equations*, American Mathematical Society, Providence, 1998.
- [3] R. B. Holmes. *A Course on Optimization and Best Approximation*. Springer-Verlag, Berlin, 1972.
- [4] Harold S. Shapiro. *Smoothing and approximation of functions*. Van Nostrand Reinhold Co., New York, 1969. Revised and expanded edition of mimeographed notes (Matscience Report No. 55), Van Nostrand Reinhold Mathematical Studies.
- [5] Harold S. Shapiro. *Topics in approximation theory*. Springer-Verlag, Berlin, 1971. With appendices by Jan Boman and Torbjörn Hedberg, Lecture Notes in Math., Vol. 187.
- [6] I. Singer. *Best Approximation in Normed Linear Spaces by Elements of Linear Subspaces*. Springer-Verlag, Berlin, 1970.