

Numerical Analysis

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with thanks to Endre Süli

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Least-Squares Approximation

Consider the problem of least-squares approximation in the inner product space $L_w^2(a, b)$, with inner product

$$\langle f, g \rangle = \int_a^b w(x) f(x) g(x) dx$$

and norm

$$\|f\|_2 = \langle f, f \rangle^{1/2}$$

where w is a weight-function, defined, positive and integrable on (a, b) .

Theorem

If $f \in L_w^2(a, b)$ and $p_n \in \Pi_n$ is such that

$$\langle f - p_n, r \rangle = 0 \quad \forall r \in \Pi_n, \quad (12.1)$$

then

$$\|f - p_n\|_2 \leq \|f - r\|_2 \quad \forall r \in \Pi_n,$$

i.e., p_n is a best (weighted) least-squares approximation to f on $[a, b]$.

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Remark: the converse is true too (see Problem Sheet 6, Q9).

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[Note that (12.2) holds if, and only if,

$$\int_a^b w(x) \left(f - \sum_{k=0}^n \alpha_k x^k \right) \left(\sum_{i=0}^n \beta_i x^i \right) dx = 0 \quad \forall q = \sum_{i=0}^n \beta_i x^i \in \Pi_n.]$$

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However, (12.2) implies that

$$\sum_{k=0}^n \left(\int_a^b w(x) x^{k+i} dx \right) \alpha_k = \int_a^b w(x) f(x) x^i dx \text{ for } i = 0, 1, \dots, n.$$

This is the component-wise statement of a matrix equation

$$A\alpha = \varphi, \tag{12.3}$$

to determine the coefficients $\alpha = (\alpha_0, \alpha_1, \dots, \alpha_n)^\top$, where $A = \{a_{i,k}, i, k = 0, 1, \dots, n\}$, $\varphi = (f_0, f_1, \dots, f_n)^\top$,

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The system (12.3) is called the set of **normal equations**.

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$$\int_0^1 [e^x - (\alpha_0 1 + \alpha_1 x)] 1 dx = 0 \quad \text{and} \quad \int_0^1 [e^x - (\alpha_0 1 + \alpha_1 x)] x dx = 0.$$

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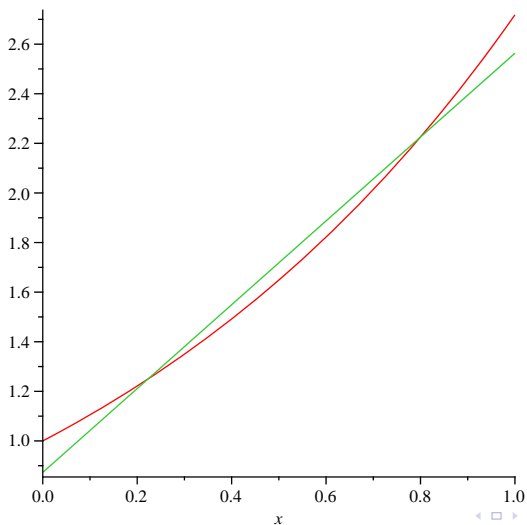
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so $p_1(x) := (18 - 6e)x + (4e - 10)$ is the best approximation.

We plot $f(x) = e^x$ and $p_1(x)$ for $x \in [0, 1]$ using Maple

```
plot( [exp(x), (18 - 6*exp(1))*x + (4*exp(1) - 10) ], x=0..1)
```



Proof that the coefficient matrix A is nonsingular will now establish existence and uniqueness of (weighted) $\|\cdot\|_2$ best-approximation.

Theorem

The coefficient matrix A is nonsingular.

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This contradicts the initial supposition, and thus A is nonsingular. □

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as a basis for Π_n , suppose we have a basis

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We shall choose the basis $\{\phi_0, \phi_1, \dots, \phi_n\}$ so as to ensure that the matrix A in the normal equations is **diagonal**.

With this choice of basis, we seek the polynomial of best approximation

$p_n(x) = \sum_{k=0}^n \beta_k \phi_k(x)$. The normal equations then become

$$\int_a^b w(x) \left(f(x) - \sum_{k=0}^n \beta_k \phi_k(x) \right) \phi_i(x) dx = 0, \quad i = 0, 1, \dots, n,$$

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i.e.,

$$A\beta = \varphi, \tag{12.4}$$

where $\beta = (\beta_0, \beta_1, \dots, \beta_n)^T$, $\varphi = (f_1, f_2, \dots, f_n)^T$ and now

$$a_{i,k} = \int_a^b w(x) \phi_k(x) \phi_i(x) dx \quad \text{and} \quad f_i = \int_a^b w(x) f(x) \phi_i(x) dx.$$

So A is diagonal if

$$\langle \phi_i, \phi_k \rangle = \int_a^b w(x) \phi_i(x) \phi_k(x) dx \quad \begin{cases} = 0 & i \neq k \text{ and} \\ \neq 0 & i = k. \end{cases}$$

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We can create such a set of **orthogonal polynomials**

$$\{\phi_0, \phi_1, \dots, \phi_n, \dots\},$$

with $\phi_i \in \Pi_i$ for each i , by the Gram–Schmidt procedure, which is based on the following lemma.

Lemma

Suppose that $\{\phi_0, \phi_1, \dots, \phi_k\}$, with $\phi_i \in \Pi_i$ for each i , are orthogonal with respect to the inner product $\langle f, g \rangle = \int_a^b w(x)f(x)g(x) dx$. Then,

$$\phi_{k+1}(x) = x^{k+1} - \sum_{i=0}^k \lambda_i \phi_i(x)$$

satisfies

$$\langle \phi_{k+1}, \phi_j \rangle = \int_a^b w(x)\phi_{k+1}(x)\phi_j(x) dx = 0, \quad j = 0, 1, \dots, k,$$

when

$$\lambda_j = \frac{\langle x^{k+1}, \phi_j \rangle}{\langle \phi_j, \phi_j \rangle}, \quad j = 0, 1, \dots, k.$$

Proof. For any j , $0 \leq j \leq k$,

$$\begin{aligned}\langle \phi_{k+1}, \phi_j \rangle &= \langle x^{k+1}, \phi_j \rangle - \sum_{i=0}^k \lambda_i \langle \phi_i, \phi_j \rangle \\ &= \langle x^{k+1}, \phi_j \rangle - \lambda_j \langle \phi_j, \phi_j \rangle\end{aligned}$$

by the orthogonality of ϕ_i and ϕ_j , $i \neq j$,

$$= 0 \quad \text{by definition of } \lambda_j. \quad \square$$

Notes:

- 1 The Gram–Schmidt procedure does the above for $k = 0, 1, \dots, n$ successively.
- 2 ϕ_k is always of exact degree k , so $\{\phi_0, \phi_1, \dots, \phi_\ell\}$ is a basis for Π_ℓ for every $\ell \geq 0$.
- 3 ϕ_k can be normalised/scaled to satisfy $\langle \phi_k, \phi_k \rangle = 1$ or to be monic.