

LEAST SQUARES APPROXIMATION

Another approach to approximating a function $f(x)$ on an interval $a \leq x \leq b$ is to seek an approximation $p(x)$ with a small ‘average error’ over the interval of approximation. A convenient definition of the average error of the approximation is given by

$$E(p; f) \equiv \left[\frac{1}{b-a} \int_a^b [f(x) - p(x)]^2 dx \right]^{\frac{1}{2}} \quad (1)$$

This is also called the *root-mean-square-error* (denoted subsequently by *RMSE*) in the approximation of $f(x)$ by $p(x)$. Note first that choosing $p(x)$ to minimize $E(p; f)$ is equivalent to minimizing

$$\int_a^b [f(x) - p(x)]^2 dx$$

thus dispensing with the square root and multiplying fraction (although the minimums are generally different). The minimizing of (1) is called the *least squares approximation problem*.

Example. Let $f(x) = e^x$, let $p(x) = \alpha_0 + \alpha_1 x$, α_0 , α_1 unknown. Approximate $f(x)$ over $[-1, 1]$. Choose α_0 , α_1 to minimize

$$g(\alpha_0, \alpha_1) \equiv \int_{-1}^1 [e^x - \alpha_0 - \alpha_1 x]^2 dx \quad (2)$$

$$g(\alpha_0, \alpha_1) = \int_{-1}^1 \left\{ \begin{array}{l} e^{2x} + \alpha_0^2 + \alpha_1^2 x^2 - 2\alpha_0 e^x \\ -2\alpha_1 x e^x + 2\alpha_0 \alpha_1 x \end{array} \right\} dx$$

Integrating,

$$g(\alpha_0, \alpha_1) = c_1 \alpha_0^2 + c_2 \alpha_1^2 + c_3 \alpha_0 \alpha_1 + c_4 \alpha_0 + c_5 \alpha_1 + c_6$$

with constants $\{c_1, \dots, c_6\}$, e.g.

$$c_1 = 2, \quad c_6 = (e^1 - e^{-1}) / 2.$$

g is a quadratic polynomial in the two variables α_0 , α_1 . To find its minimum, solve the system

$$\frac{\partial g}{\partial \alpha_0} = 0, \quad \frac{\partial g}{\partial \alpha_1} = 0$$

It is simpler to return to (2) to differentiate, obtaining

$$2 \int_{-1}^1 [e^x - \alpha_0 - \alpha_1 x] (-1) dx = 0$$

$$2 \int_{-1}^1 [e^x - \alpha_0 - \alpha_1 x] (-x) dx = 0$$

This simplifies to

$$2\alpha_0 = \int_{-1}^1 e^x dx = e - e^{-1}$$

$$\frac{2}{3}\alpha_1 = \int_{-1}^1 x e^x dx = 2e^{-1}$$

$$\alpha_0 = \frac{e - e^{-1}}{2} \doteq 1.1752$$

$$\alpha_1 = 3e^{-1} \doteq 1.1036$$

Using these values for α_0 and α_1 , we denote the resulting linear approximation by

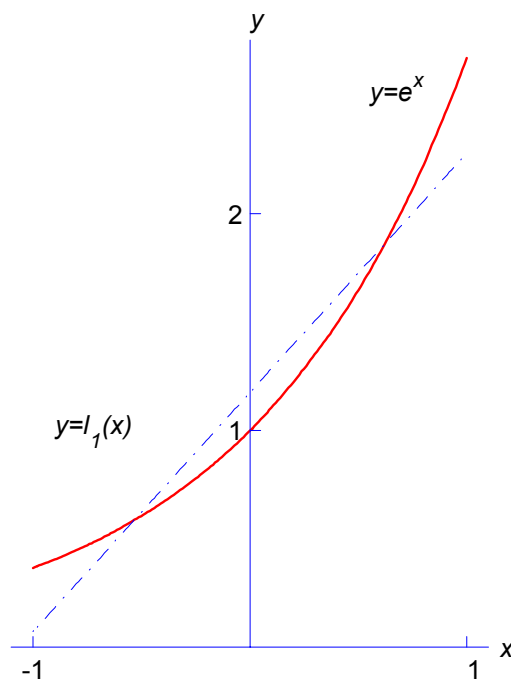
$$\ell_1(x) = \alpha_0 + \alpha_1 x$$

It is called the best linear approximation to e^x in the *sense of least squares*. For the error,

$$\max_{-1 \leq x \leq 1} |e^x - \ell_1(x)| \doteq 0.439$$

Errors in linear approximations of e^x :

<i>Approximation</i>	<i>Max Error</i>	<i>RMSE</i>
Taylor $t_1(x)$	0.718	0.246
Least squares $l_1(x)$	0.439	0.162
Chebyshev $c_1(x)$	0.372	0.184
Minimax $m_1(x)$	0.279	0.190



The linear least squares approximation to e^x

THE GENERAL CASE

Approximate $f(x)$ on $[a, b]$, and let $n \geq 0$. Seek $p(x)$ to minimize the *RMSE*. Write

$$p(x) = \alpha_0 + \alpha_1 x + \cdots + \alpha_n x^n$$

$$g(\alpha_0, \alpha_1, \dots, \alpha_n) \equiv \int_{-1}^1 \left[\begin{array}{c} f(x) - \alpha_0 - \alpha_1 x \\ \quad \quad \quad - \cdots - \alpha_n x^n \end{array} \right]^2 dx$$

Find coefficients $\alpha_0, \alpha_1, \dots, \alpha_n$ to minimize this integral. The integral $g(\alpha_0, \alpha_1, \dots, \alpha_n)$ is a quadratic polynomial in the $n + 1$ variables $\alpha_0, \alpha_1, \dots, \alpha_n$.

To minimize $g(\alpha_0, \alpha_1, \dots, \alpha_n)$, invoke the conditions

$$\frac{\partial g}{\partial \alpha_i} = 0, \quad i = 0, 1, \dots, n$$

This yields a set of $n + 1$ equations that must be satisfied by a minimizing set $\alpha_0, \alpha_1, \dots, \alpha_n$ for g . Manipulating this set of conditions leads to a simultaneous linear system.

To better understand the form of the linear system, consider the special case of $[a, b] = [0, 1]$. Differentiating g with respect to each α_i , we obtain

$$\begin{aligned} 2 \int_{-1}^1 [e^x - \alpha_0 - \cdots - \alpha_n x^n] (-1) dx &= 0 \\ 2 \int_{-1}^1 [e^x - \alpha_0 - \cdots - \alpha_n x^n] (-x) dx &= 0 \\ &\vdots \\ 2 \int_{-1}^1 [e^x - \alpha_0 - \cdots - \alpha_n x^n] (-x^n) dx &= 0 \end{aligned}$$

Then the linear system is

$$\sum_{j=0}^n \frac{\alpha_j}{i+j+1} = \int_0^1 x^i f(x) dx, \quad i = 0, 1, \dots, n$$

We will study the solution of simultaneous linear systems in Chapter 6. There we will see that this linear system is ‘ill-conditioned’ and is difficult to solve accurately, even for moderately sized values of n such as $n = 5$. As a consequence, this is not a good approach to solving for a minimizer of $g(\alpha_0, \alpha_1, \dots, \alpha_n)$.

LEGENDRE POLYNOMIALS

Define the *Legendre polynomials* as follows.

$$P_0(x) = 1$$

$$P_n(x) = \frac{1}{n!2^n} \cdot \frac{d^n}{dx^n} \left[(x^2 - 1)^n \right], \quad n = 1, 2, \dots$$

For example,

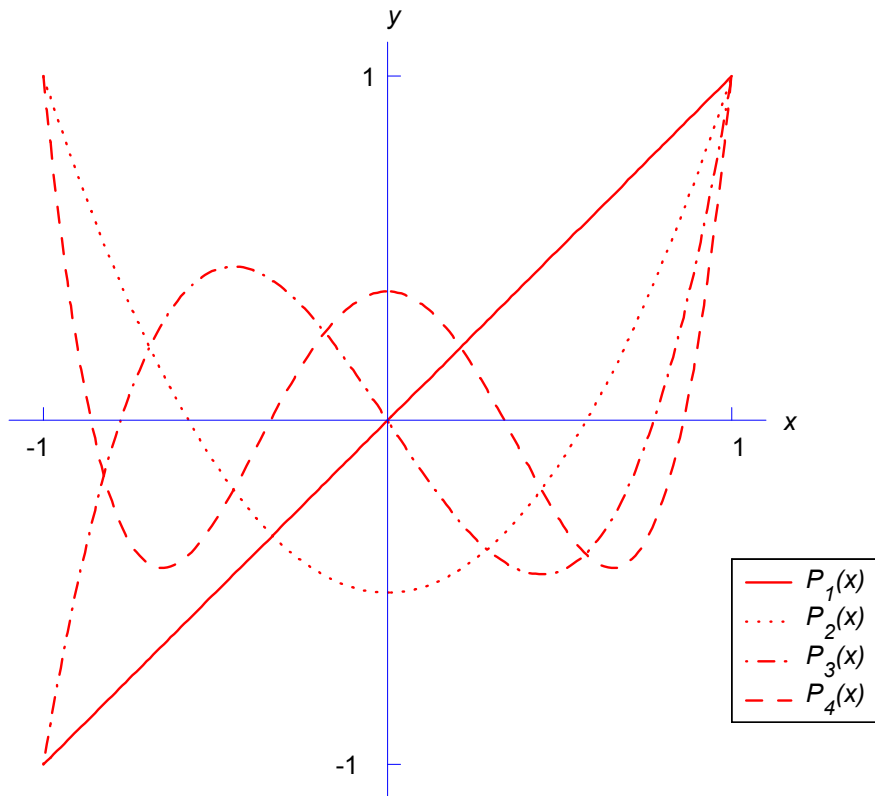
$$P_1(x) = x$$

$$P_2(x) = \frac{1}{2} (3x^2 - 1)$$

$$P_3(x) = \frac{1}{2} (5x^3 - 3x)$$

$$P_4(x) = \frac{1}{8} (35x^4 - 30x^2 + 3)$$

The Legendre polynomials have many special properties, and they are widely used in numerical analysis and applied mathematics.



Legendre polynomials of degrees 1, 2, 3, 4

PROPERTIES

Introduce the special notation

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

for general functions $f(x)$ and $g(x)$.

- *Degree and normalization:*

$$\deg P_n = n, \quad P_n(1) = 1, \quad n \geq 0$$

- *Triple recursion relation:* For $n \geq 1$,

$$P_{n+1}(x) = \frac{2n+1}{n+1}xP_n(x) - \frac{n}{n+1}P_{n-1}(x)$$

- *Orthogonality and size:*

$$(P_i, P_j) = \begin{cases} 0, & i \neq j \\ \frac{2}{2j+1}, & i = j \end{cases}$$

- *Zeroes:*

All zeroes of $P_n(x)$ are located in $[-1, 1]$;
all zeroes are simple roots of $P_n(x)$

- *Basis:* Every polynomial $p(x)$ of degree $\leq n$ can be written in the form

$$p(x) = \sum_{j=0}^n \beta_j P_j(x)$$

with the choice of $\beta_0, \beta_1, \dots, \beta_n$ uniquely determined from $p(x)$:

$$\beta_j = \frac{(p, P_j)}{(P_j, P_j)}, \quad j = 0, 1, \dots, n$$

FINDING THE LEAST SQUARES APPROXIMATION

We solve the least squares approximation problem on only the interval $[-1, 1]$. Approximation problems on other intervals $[a, b]$ can be accomplished using a linear change of variable.

We seek to find a polynomial $p(x)$ of degree n that minimizes

$$\int_a^b [f(x) - p(x)]^2 dx$$

This is equivalent to minimizing

$$(f - p, f - p) \tag{3}$$

We begin by writing $p(x)$ in the form

$$p(x) = \sum_{j=0}^n \beta_j P_j(x)$$

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Substitute into (3), obtaining

$$\begin{aligned} \tilde{g}(\beta_0, \beta_1, \dots, \beta_n) &\equiv (f - p, f - p) \\ &= \left(f - \sum_{j=0}^n \beta_j P_j, f - \sum_{i=0}^n \beta_i P_i \right) \end{aligned}$$

Expand this into the following:

$$\begin{aligned} \tilde{g} &= (f, f) - \sum_{j=0}^n \frac{(f, P_j)^2}{(P_j, P_j)} \\ &\quad + \sum_{j=0}^n (P_j, P_j) \left[\beta_j - \frac{(f, P_j)}{(P_j, P_j)} \right]^2 \end{aligned}$$

Looking at this carefully, we see that it is smallest when

$$\beta_j = \frac{(f, P_j)}{(P_j, P_j)}, \quad j = 0, 1, \dots, n$$

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The minimum for this choice of coefficients is

$$\tilde{g} = (f, f) - \sum_{j=0}^n \frac{(f, P_j)^2}{(P_j, P_j)}$$

We call

$$\ell_n(x) = \sum_{j=0}^n \frac{(f, P_j)}{(P_j, P_j)} P_j(x) \quad (4)$$

the *least squares approximation of degree n* to $f(x)$ on $[-1, 1]$.

If $\beta_n = 0$, then its actual degree is less than n .

Example. Approximate $f(x) = e^x$ on $[-1, 1]$. We use (4) with $n = 3$:

$$\ell_3(x) = \sum_{j=0}^3 \beta_j P_j(x), \quad \beta_j = \frac{(f, P_j)}{(P_j, P_j)} \quad (5)$$

The coefficients $\{\beta_0, \beta_1, \beta_2, \beta_3\}$ are as follows.

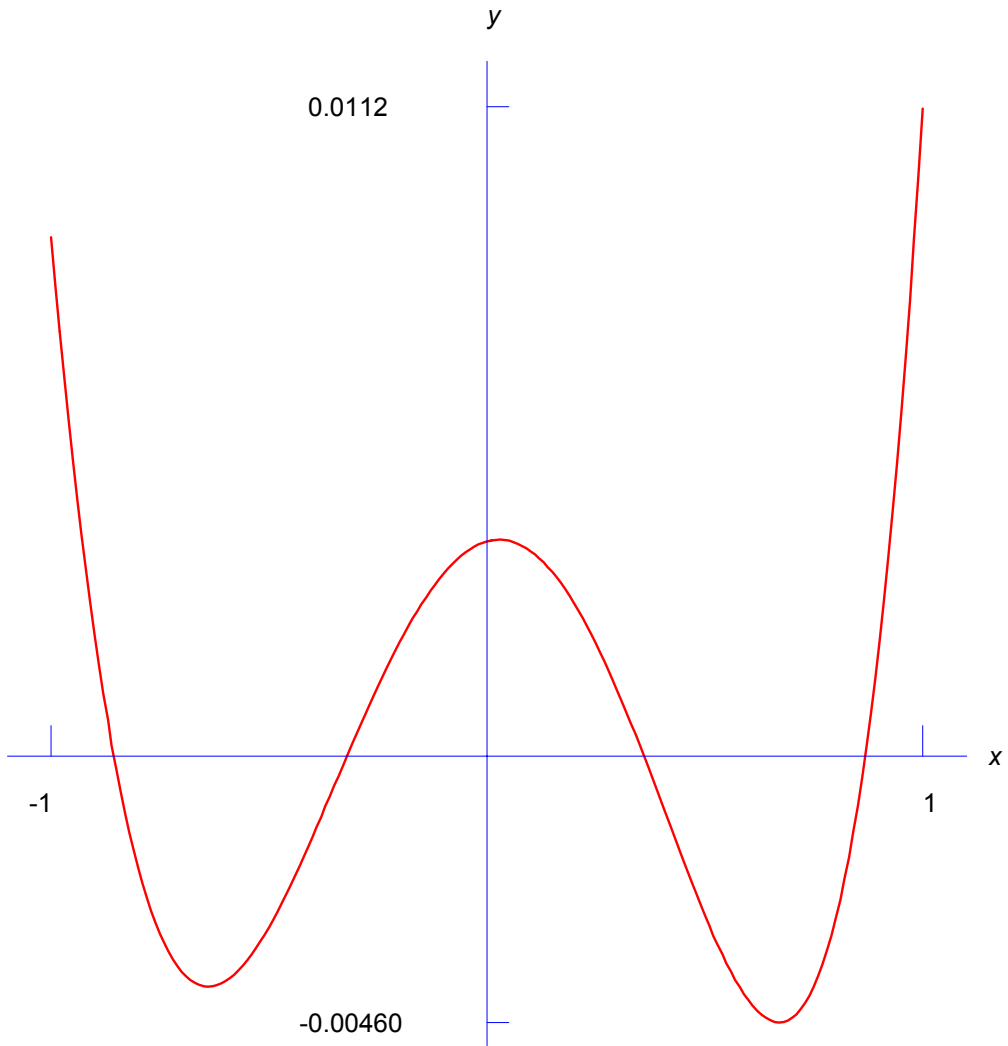
j	0	1	2	3
β_j	2.35040	0.73576	0.14313	0.02013

Using (5) and simplifying,

$$\ell_3(x) = .996294 + .997955x + .536722x^2 + .176139x^3$$

The error in various cubic approximations:

<i>Approximation</i>	<i>Max Error</i>	<i>RMSE</i>
Taylor $t_3(x)$.0516	.0145
Least squares $\ell_3(x)$.0112	.00334
Chebyshev $c_3(x)$.00666	.00384
Minimax $m_3(x)$.00553	.00388



Error in the cubic least squares approximation to e^x