

## 4.4 Best Approximation

Given a function  $f(x)$  that is continuous on a given interval  $[a, b]$ , consider approximating it by some polynomial  $p(x)$ . To measure the error in  $p(x)$  as an approximation, introduce

$$E(p) = \max_{a \leq x \leq b} |f(x) - p(x)|$$

This is called the *maximum error* or *uniform error* of approximation of  $f(x)$  by  $p(x)$  on  $[a, b]$ .

With an eye towards efficiency, we want to find the 'best' possible approximation of a given degree  $n$ . With this in mind, introduce the following:

$$\begin{aligned}\rho_n(f) &= \min_{\deg(p) \leq n} E(p) \\ &= \min_{\deg(p) \leq n} \left[ \max_{a \leq x \leq b} |f(x) - p(x)| \right]\end{aligned}$$

The number  $\rho_n(f)$  will be the smallest possible uniform error, or *minimax error*, when approximating  $f(x)$  by polynomials of degree at most  $n$ . If there is a polynomial giving this smallest error, we denote it by  $m_n(x)$ ; thus  $E(m_n) = \rho_n(f)$ .

$$E(p) = \max_{a \leq x \leq b} |f(x) - p(x)|$$

**Example.** Let  $f(x) = e^x$  on  $[-1, 1]$ . In the following table, we give the values of  $E(t_n)$ ,  $t_n(x)$  the Taylor polynomial of degree  $n$  for  $e^x$  about  $x = 0$ , and  $E(m_n)$ .

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	Maximum Error in:	
$n$	$t_n(x)$	$m_n(x)$
1	7.18E - 1	2.79E - 1
2	2.18E - 1	4.50E - 2
3	5.16E - 2	5.53E - 3
4	9.95E - 3	5.47E - 4
5	1.62E - 3	4.52E - 5
6	2.26E - 4	3.21E - 6
7	2.79E - 5	2.00E - 7
8	3.06E - 6	1.11E - 8
9	3.01E - 7	5.52E - 10

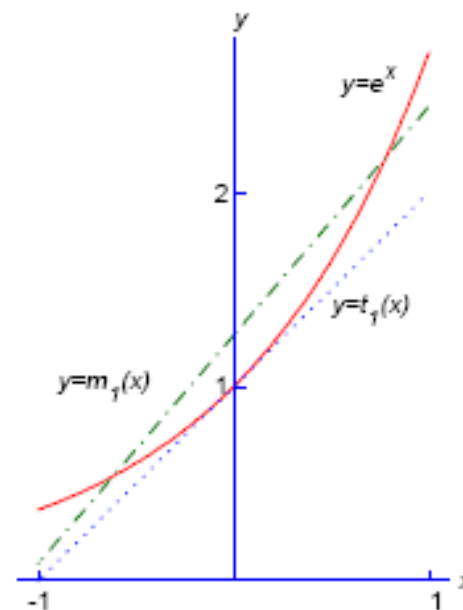
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Consider graphically how we can improve on the Taylor polynomial  $t_1(x) = 1 + x$

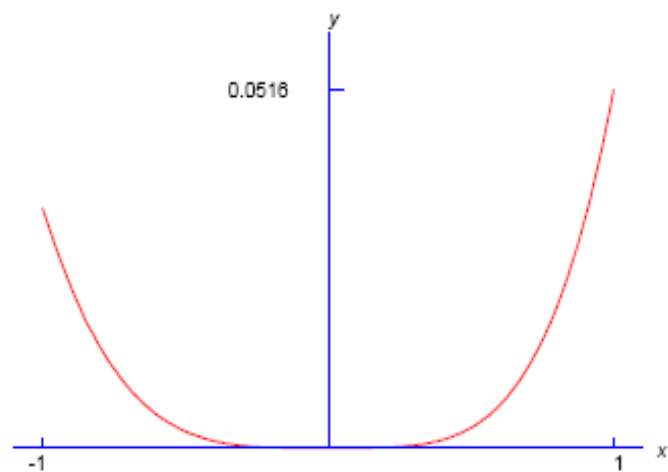
as a uniform approximation to  $e^x$  on the interval  $[-1, 1]$ .

The linear minimax approximation is

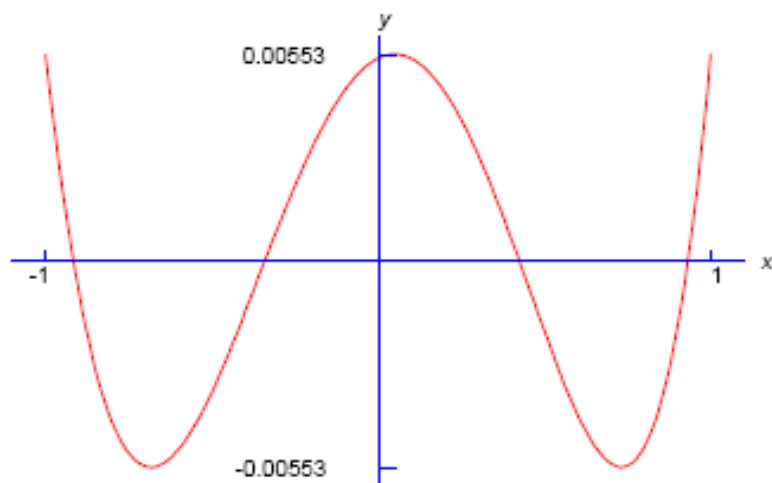
$$m_1(x) = 1.2643 + 1.1752x$$



Linear Taylor and minimax approximations to  $e^x$



Error in cubic Taylor approximation to  $e^x$



Error in cubic minimax approximation to  $e^x$

## Accuracy of the minimax approximation.

$$\rho_n(f) \leq \frac{[(b-a)/2]^{n+1}}{(n+1)!2^n} \max_{a \leq x \leq b} |f^{(n+1)}(x)|$$

This error bound does not always become smaller with increasing  $n$ , but it will give a fairly accurate bound for many common functions  $f(x)$ .

**Example.** Let  $f(x) = e^x$  for  $-1 \leq x \leq 1$ . Then

$$\rho_n(e^x) \leq \frac{e}{(n+1)!2^n} \quad (*)$$

$n$	$Bound (*)$	$\rho_n(f)$
1	6.80E - 1	2.79E - 1
2	1.13E - 1	4.50E - 2
3	1.42E - 2	5.53E - 3
4	1.42E - 3	5.47E - 4
5	1.18E - 4	4.52E - 5
6	8.43E - 6	3.21E - 6
7	5.27E - 7	2.00E - 7

## 4.5 Chebyshev Polynomials

*Chebyshev polynomials* are used in many parts of numerical analysis, and more generally, in applications of mathematics. For an integer  $n \geq 0$ , define the function

$$T_n(x) = \cos \left( n \cos^{-1} x \right), \quad -1 \leq x \leq 1 \quad (1)$$

This may not appear to be a polynomial, but we will show it is a polynomial of degree  $n$ . To simplify the manipulation of (1), we introduce

$$\theta = \cos^{-1}(x) \quad \text{or} \quad x = \cos(\theta), \quad 0 \leq \theta \leq \pi \quad (2)$$



$$\theta = \cos^{-1}(x) \quad \text{or} \quad x = \cos(\theta), \quad 0 \leq \theta \leq \pi \quad (2)$$

Then

$$T_n(x) = \cos(n\theta) \quad (3)$$

**Example.**  $n = 0$

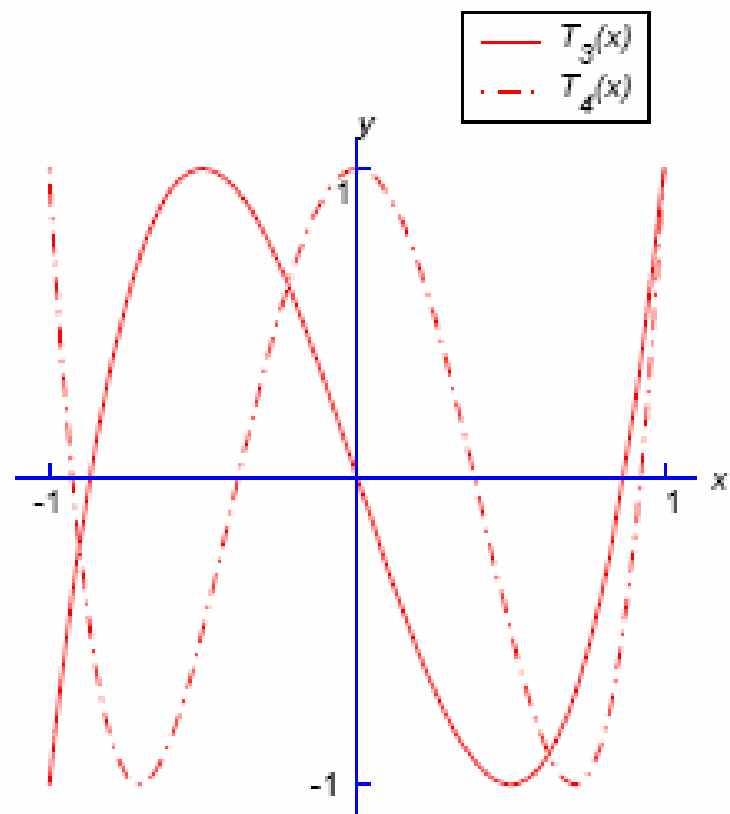
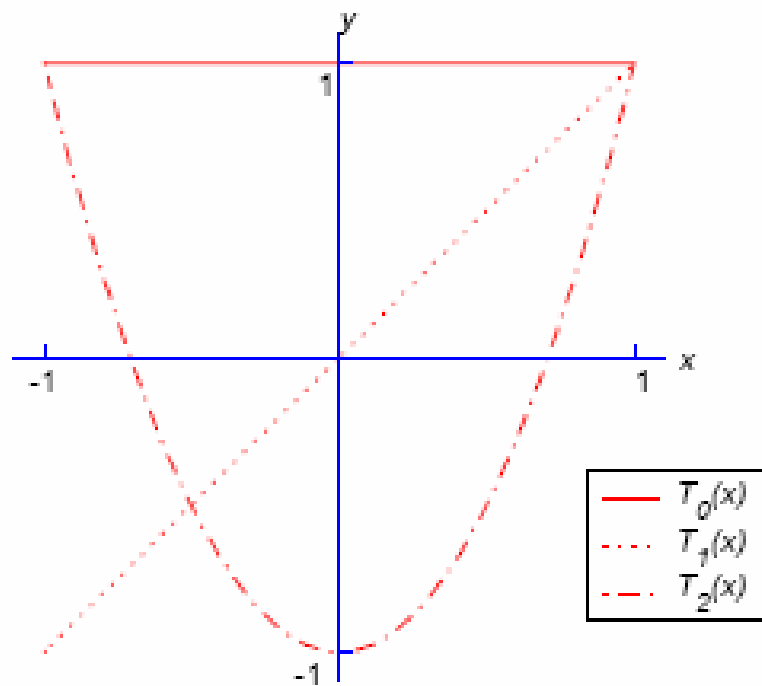
$$T_0(x) = \cos(0 \cdot \theta) = 1$$

$$n = 1$$

$$T_1(x) = \cos(\theta) = x$$

$$n = 2$$

$$T_2(x) = \cos(2\theta) = 2 \cos^2(\theta) - 1 = 2x^2 - 1$$



**The triple recursion relation.** Recall the trigonometric addition formulas,

$$\cos(\alpha \pm \beta) = \cos(\alpha) \cos(\beta) \mp \sin(\alpha) \sin(\beta)$$

Let  $n \geq 1$ , and apply these identities to get

$$\begin{aligned} T_{n+1}(x) &= \cos[(n+1)\theta] = \cos(n\theta + \theta) \\ &= \cos(n\theta) \cos(\theta) - \sin(n\theta) \sin(\theta) \end{aligned}$$

$$\begin{aligned} T_{n-1}(x) &= \cos[(n-1)\theta] = \cos(n\theta - \theta) \\ &= \cos(n\theta) \cos(\theta) + \sin(n\theta) \sin(\theta) \end{aligned}$$

Add these two equations, and then use (1) and (3) to obtain

$$\begin{aligned}T_{n+1}(x) + T_{n-1} &= 2 \cos(n\theta) \cos(\theta) = 2xT_n(x) \\T_{n+1}(x) &= 2xT_n(x) - T_{n-1}(x), \quad n \geq 1\end{aligned}\tag{4}$$

This is called the *triple recursion relation* for the Chebyshev polynomials. It is often used in evaluating them, rather than using the explicit formula (1).

**Example.** Recall

$$T_0(x) = 1, \quad T_1(x) = x$$

$$T_{n+1}(x) = 2xT_n(x) - T_{n-1}(x), \quad n \geq 1$$

Let  $n = 2$ . Then

$$\begin{aligned}T_3(x) &= 2xT_2(x) - T_1(x) \\ &= 2x(2x^2 - 1) - x \\ &= 4x^3 - 3x\end{aligned}$$

Let  $n = 3$ . Then

$$\begin{aligned}T_4(x) &= 2xT_3(x) - T_2(x) \\ &= 2x(4x^3 - 3x) - (2x^2 - 1) \\ &= 8x^4 - 8x^2 + 1\end{aligned}$$

**The minimum size property.** Note that

$$|T_n(x)| \leq 1, \quad -1 \leq x \leq 1 \quad (5)$$

for all  $n \geq 0$ . Also, note that

$$T_n(x) = 2^{n-1}x^n + \text{lower degree terms}, \quad n \geq 1 \quad (6)$$

This can be proven using the triple recursion relation and mathematical induction.

Introduce a modified version of  $T_n(x)$ ,

$$\tilde{T}_n(x) = \frac{1}{2^{n-1}}T_n(x) = x^n + \text{lower degree terms} \quad (7)$$

From  $|T_n(x)| \leq 1$ , and  $T_n(x) = 2^{n-1}x^n + \text{lower degree terms}$ ,  
we get

$$|\tilde{T}_n(x)| \leq \frac{1}{2^{n-1}}, \quad -1 \leq x \leq 1, \quad n \geq 1 \quad (8)$$

**Example.**

$$\tilde{T}_4(x) = \frac{1}{8} (8x^4 - 8x^2 + 1) = x^4 - x^2 + \frac{1}{8}$$

A polynomial whose highest degree term has a coefficient of 1 is called a *monic polynomial*. Formula (8) says the monic polynomial  $\tilde{T}_n(x)$  has size  $1/2^{n-1}$  on  $-1 \leq x \leq 1$ , and this becomes smaller as the degree  $n$  increases. In comparison,

$$\max_{-1 \leq x \leq 1} |x^n| = 1$$

Thus  $x^n$  is a monic polynomial whose size does not change with increasing  $n$ .



**Theorem.** Let  $n \geq 1$  be an integer, and consider all possible monic polynomials of degree  $n$ . Then the degree  $n$  monic polynomial with the smallest maximum on  $[-1, 1]$  is the modified Chebyshev polynomial  $\tilde{T}_n(x)$ , and its maximum value on  $[-1, 1]$  is  $1/2^{n-1}$ .

This result is used in devising applications of Chebyshev polynomials. We apply it to obtain an improved interpolation scheme.

## 4.6 A Near-Minimax Approximation

Let  $f(x)$  be continuous on  $[a, b] = [-1, 1]$ . Consider approximating  $f$  by an interpolatory polynomial of degree at most  $n = 3$ . Let  $x_0, x_1, x_2, x_3$  be interpolation node points in  $[-1, 1]$ ; let  $c_3(x)$  be of degree  $\leq 3$  and interpolate  $f(x)$  at  $\{x_0, x_1, x_2, x_3\}$ . The interpolation error is

$$f(x) - c_3(x) = \frac{\omega(x)}{4!} f^{(4)}(\xi_x), \quad -1 \leq x \leq 1 \quad (1)$$

$$\omega(x) = (x - x_0)(x - x_1)(x - x_2)(x - x_3) \quad (2)$$

with  $\xi_x$  in  $[-1, 1]$ . We want to choose the nodes  $\{x_0, x_1, x_2, x_3\}$  so as to minimize the maximum value of  $|f(x) - c_3(x)|$  on  $[-1, 1]$ .

$$f(x) - c_3(x) = \frac{\omega(x)}{4!} f^{(4)}(\xi_x), \quad -1 \leq x \leq 1 \quad (1)$$

$$\omega(x) = (x - x_0)(x - x_1)(x - x_2)(x - x_3) \quad (2)$$

From (1), the only general quantity, independent of  $f$ , is  $\omega(x)$ . Thus we choose  $\{x_0, x_1, x_2, x_3\}$  to minimize

$$\max_{-1 \leq x \leq 1} |\omega(x)| \quad (3)$$

Expand to get

$$\omega(x) = x^4 + \text{lower degree terms}$$

This is a monic polynomial of degree 4. From the theorem in the preceding section, the smallest possible value for (3) is obtained with

$$\omega(x) = \tilde{T}_4(x) = \frac{T_4(x)}{2^3} = \frac{1}{8}(8x^4 - 8x^2 + 1) \quad (4)$$

$$\max_{-1 \leq x \leq 1} |\omega(x)| \quad (3)$$

and the smallest value of (3) is  $1/2^3$  in this case. The equation (4) defines implicitly the nodes  $\{x_0, x_1, x_2, x_3\}$ : they are the roots of  $T_4(x)$ .

In our case this means solving

$$T_4(x) = \cos(4\theta) = 0, \quad x = \cos(\theta)$$

$$4\theta = \pm \frac{\pi}{2}, \pm \frac{3\pi}{2}, \pm \frac{5\pi}{2}, \pm \frac{7\pi}{2}, \dots$$

$$\theta = \pm \frac{\pi}{8}, \pm \frac{3\pi}{8}, \pm \frac{5\pi}{8}, \pm \frac{7\pi}{8}, \dots$$

$$x = \cos\left(\frac{\pi}{8}\right), \cos\left(\frac{3\pi}{8}\right), \cos\left(\frac{5\pi}{8}\right), \dots \quad (5)$$

using  $\cos(-\theta) = \cos(\theta)$ .

$$x = \cos\left(\frac{\pi}{8}\right), \cos\left(\frac{3\pi}{8}\right), \cos\left(\frac{5\pi}{8}\right), \cos\left(\frac{7\pi}{8}\right), \dots$$

The first four values are distinct; the following ones are repetitive. For example,

$$\cos\left(\frac{9\pi}{8}\right) = \cos\left(\frac{7\pi}{8}\right)$$

The first four values are

$$\{x_0, x_1, x_2, x_3\} = \{\pm 0.382683, \pm 0.923880\} \quad (6)$$

**Example.** Let  $f(x) = e^x$  on  $[-1, 1]$ . Use these nodes to produce the interpolating polynomial  $c_3(x)$  of degree 3. From the interpolation error formula and the bound of  $1/2^3$  for  $|\omega(x)|$  on  $[-1, 1]$ , we have

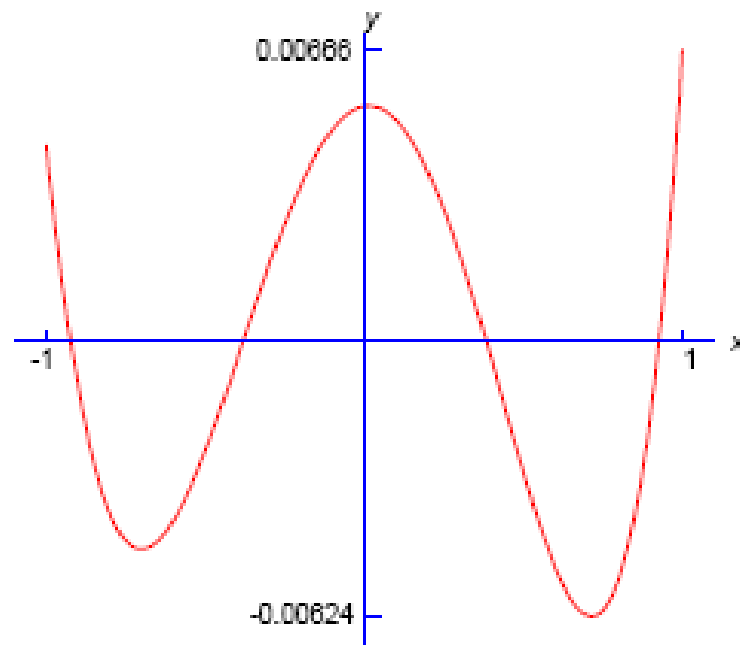
$$\begin{aligned} \max_{-1 \leq x \leq 1} |f(x) - c_3(x)| &\leq \frac{1/2^3}{4!} \max_{-1 \leq x \leq 1} e^{\xi_x} \\ &\leq \frac{e}{192} \doteq 0.014158 \end{aligned}$$

By direct calculation,

$$\max_{-1 \leq x \leq 1} |e^x - c_3(x)| \doteq 0.00666$$

### Interpolation Data: $f(x) = e^x$

$i$	$x_i$	$f(x_i)$	$f[x_0, \dots, x_i]$
0	0.923880	2.5190442	2.5190442
1	0.382683	1.4662138	1.9453769
2	-0.382683	0.6820288	0.7047420
3	-0.923880	0.3969760	0.1751757



The error  $e^x - c_3(x)$

For comparison,  $E(t_3) \doteq 0.0142$  and  $\rho_3(e^x) \doteq 0.00553$ .

For  $f(x) = e^x$ , find  $c_1(x)$ .

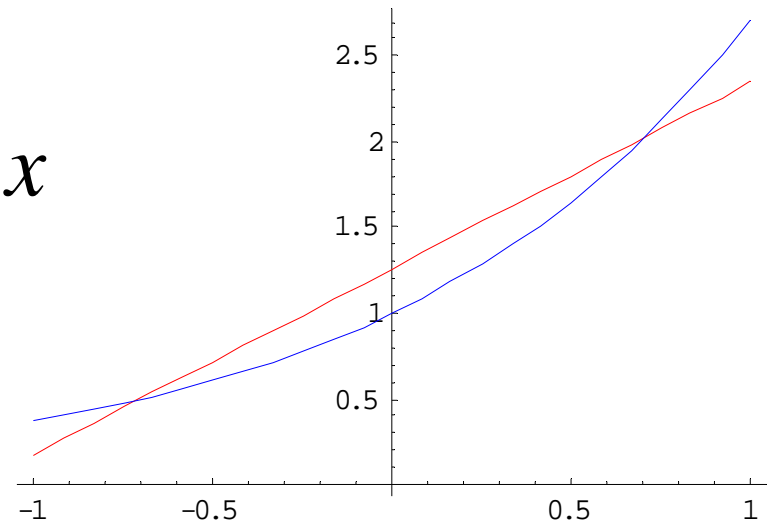
In this case we have

$$w(x) = (x - x_0)(x - x_1) = \frac{T_2(x)}{2} = x^2 - \frac{1}{2}$$

Hence  $x_0 = -1/\sqrt{2} = -0.707107$  and  $x_1 = 1/\sqrt{2} = 0.707107$

But then  $y_0 = e^{-0.707107} = 0.493069$  and  $y_1 = e^{0.707107} = 2.02811$

$$c_1(x) = 1.26059 + 1.08544x$$





## THE GENERAL CASE

Consider interpolating  $f(x)$  on  $[-1, 1]$  by a polynomial of degree  $\leq n$ , with the interpolation nodes  $\{x_0, \dots, x_n\}$  in  $[-1, 1]$ . Denote the interpolation polynomial by  $c_n(x)$ . The interpolation error on  $[-1, 1]$  is given by

$$\begin{aligned} f(x) - c_n(x) &= \frac{\omega(x)}{(n+1)!} f^{(n+1)}(\xi_x) & (7) \\ \omega(x) &= (x - x_0) \cdots (x - x_n) \end{aligned}$$

with  $\xi_x$  and unknown point in  $[-1, 1]$ . In order to minimize the interpolation error, we seek to minimize

$$\max_{-1 \leq x \leq 1} |\omega(x)| \quad (8)$$

The polynomial being minimized is monic of degree  $n + 1$ ,

$$\omega(x) = x^{n+1} + \text{lower degree terms}$$

From the theorem of the preceding section, this minimum is attained by the monic polynomial

$$\tilde{T}_{n+1}(x) = \frac{1}{2^n} T_{n+1}(x)$$

Thus the interpolation nodes are the zeros of  $T_{n+1}(x)$ ; and by the procedure that led to (5), they are given by

$$x_j = \cos\left(\frac{2j+1}{2n+2}\pi\right), \quad j = 0, 1, \dots, n \quad (9)$$

The near-minimax approximation  $c_n(x)$  of degree  $n$  is obtained by interpolating to  $f(x)$  at these  $n+1$  nodes on  $[-1, 1]$ .

The polynomial  $c_n(x)$  is sometimes called a *Chebyshev approximation*.

**Example.** Let  $f(x) = e^x$ . the following table contains the maximum errors in  $c_n(x)$  on  $[-1, 1]$  for varying  $n$ . For comparison, we also include the corresponding minimax errors. These figures illustrate that for practical purposes,  $c_n(x)$  is a satisfactory replacement for the minimax approximation  $m_n(x)$ .

$n$	$\max  e^x - c_n(x) $	$\rho_n(e^x)$
1	3.72E - 1	2.79E - 1
2	5.65E - 2	4.50E - 2
3	6.66E - 3	5.53E - 3
4	6.40E - 4	5.47E - 4
5	5.18E - 5	4.52E - 5
6	3.80E - 6	3.21E - 6

## THEORETICAL INTERPOLATION ERROR

For the error

$$f(x) - c_n(x) = \frac{\omega(x)}{(n+1)!} f^{(n+1)}(\xi_x)$$

we have

$$\max_{-1 \leq x \leq 1} |f(x) - c_n(x)| \leq \frac{\max_{-1 \leq x \leq 1} |\omega(x)|}{(n+1)!} \max_{-1 \leq \xi \leq 1} |f(\xi)|$$

From the theorem of the preceding section,

$$\max_{-1 \leq x \leq 1} |\tilde{T}_{n+1}(x)| = \max_{-1 \leq x \leq 1} |\omega(x)| = \frac{1}{2^n}$$

in this case. Thus

$$\max_{-1 \leq x \leq 1} |f(x) - c_n(x)| \leq \frac{1}{(n+1)! 2^n} \max_{-1 \leq \xi \leq 1} |f(\xi)|$$

## OTHER INTERVALS

Consider approximating  $f(x)$  on the finite interval  $[a, b]$ . Introduce the linear change of variables

$$x = \frac{1}{2} [(1-t)a + (1+t)b] \quad (10)$$

$$t = \frac{2}{b-a} \left[ x - \frac{b+a}{2} \right] \quad (11)$$

Introduce

$$F(t) = f \left( \frac{1}{2} [(1-t)a + (1+t)b] \right), \quad -1 \leq t \leq 1$$

The function  $F(t)$  on  $[-1, 1]$  is equivalent to  $f(x)$  on  $[a, b]$ , and we can move between them via (10)-(11). We can now proceed to approximate  $f(x)$  on  $[a, b]$  by instead approximating  $F(t)$  on  $[-1, 1]$ .

**Example.** Approximating  $f(x) = \cos x$  on  $[0, \pi/2]$  is equivalent to approximating

$$F(t) = \cos\left(\frac{1+t}{4}\pi\right), \quad -1 \leq t \leq 1$$

## 4.7 Least Square 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

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

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  $p(x) = \alpha_0 + \alpha_1 x$ ,  $\alpha_0$ ,  $\alpha_1$  unknown. Approximate  $f(x)$  over  $[-1, 1]$ .

Choose  $\alpha_0$ ,  $\alpha_1$  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.

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

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

$g$  is a quadratic polynomial in the two variables  $\alpha_0$   $\alpha_1$ .

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,

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

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

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

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

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

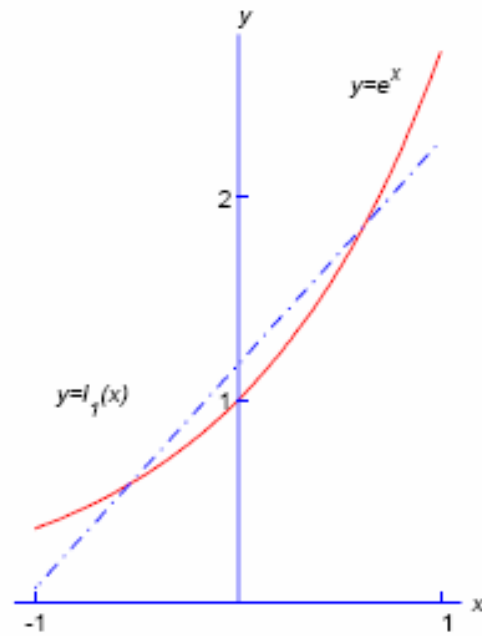
$$\alpha_1 = 3e^{-1} \doteq 1.1036$$

Using these values for  $\alpha_0$  and  $\alpha_1$ , we denote the resulting linear approximation by

$$l_1(x) = \alpha_0 + \alpha_1 x$$

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

$$\max_{-1 \leq x \leq 1} |e^x - l_1(x)| \doteq 0.439$$



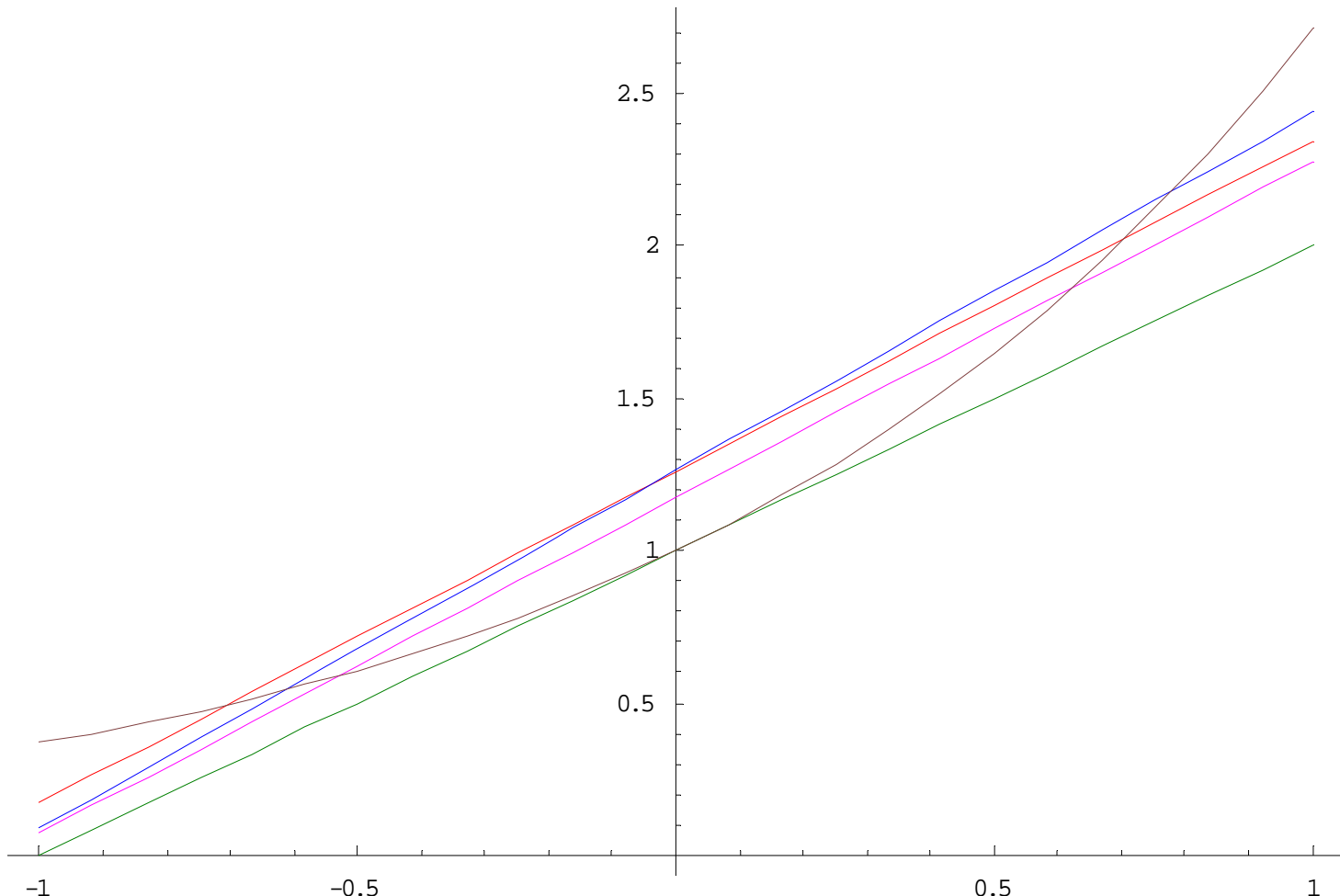
The linear least squares approximation to  $e^x$

$$t_1(x) = 1 + x$$

$$m_1(x) = 1.2643 + 1.1752x$$

$$c_1(x) = 1.26059 + 1.08544x$$

$$l_1(x) = 1.1752 + 1.1036x$$



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 $\ell_1(x)$	0.439	0.162
Chebyshev $c_1(x)$	0.372	0.184
Minimax $m_1(x)$	0.279	0.190

RMSE= Root Mean Square Error

$$E(p; f) = \sqrt{\frac{1}{b-a} \int_a^b [f(x) - p(x)]^2 dx}$$

## 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  $\alpha_i$ , 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

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

## LEGENDRE POLYNOMIALS

Define the *Legendre polynomials* as follows.

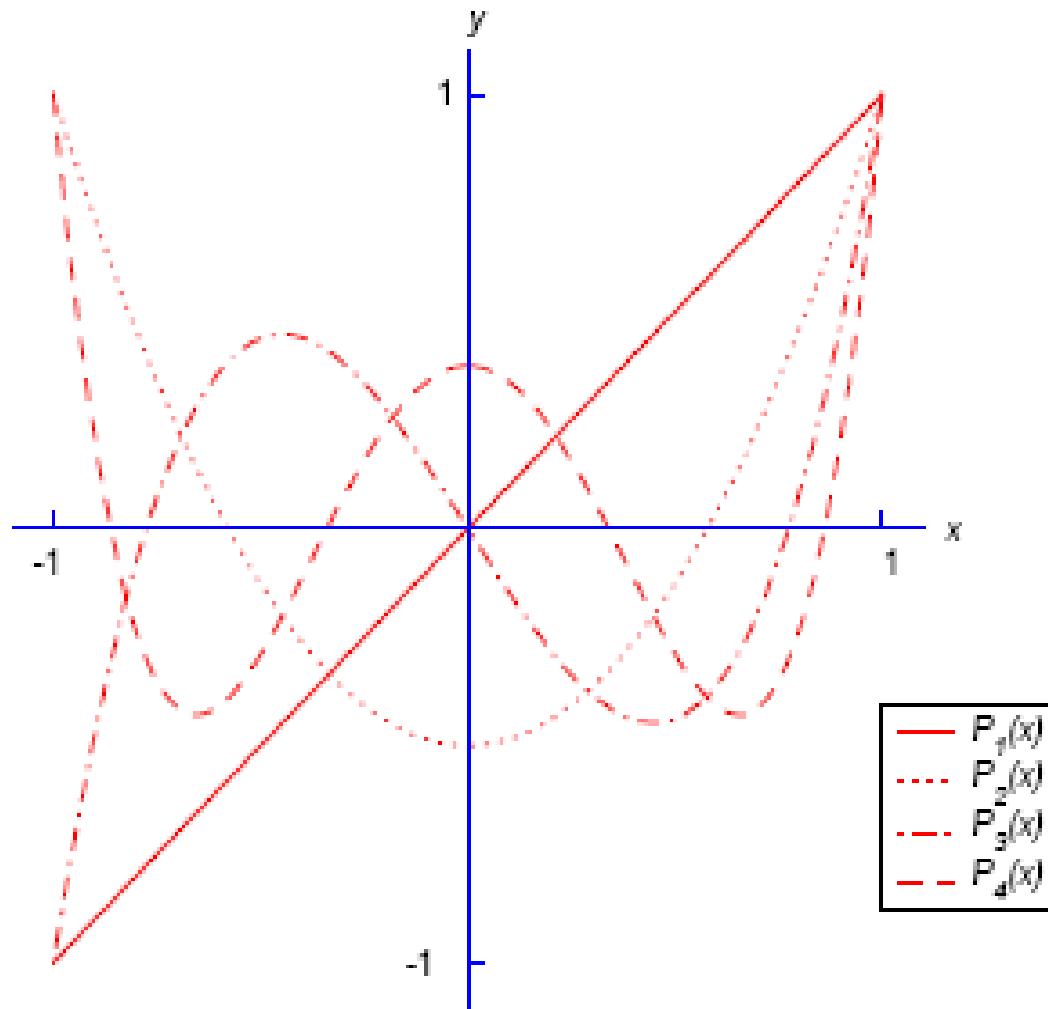
$$P_0(x) = 1$$

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

For example,

$$P_1(x) = x \qquad P_2(x) = \frac{1}{2} (3x^2 - 1)$$

$$P_3(x) = \frac{1}{2} (5x^3 - 3x) \qquad P_4(x) = \frac{1}{8} (35x^4 - 30x^2 + 3)$$



Legendre polynomials of degrees 1, 2, 3, 4

## PROPERTIES

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

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

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

Substitute into (3)  $(f - p, f - p)$

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



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


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

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

$$p(x) = \sum_{j=0}^n \beta_j P_j(x) \quad \text{minimizes} \quad \int_a^b [f(x) - p(x)]^2 dx$$


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


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


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

We call

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

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

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

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

**Example.** Approximate  $f(x) = e^x$  on  $[-1, 1]$ .

We use  $l_n(x) = \sum_{j=0}^n \frac{(f, P_j)}{(P_j, P_j)} P_j(x)$  with  $n = 3$ :

$$l_3(x) = \sum_{j=0}^3 \beta_j P_j(x), \quad \beta_j = \frac{(f, P_j)}{(P_j, P_j)}$$

$$P_0(x) = 1$$

$$P_1(x) = x$$

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

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

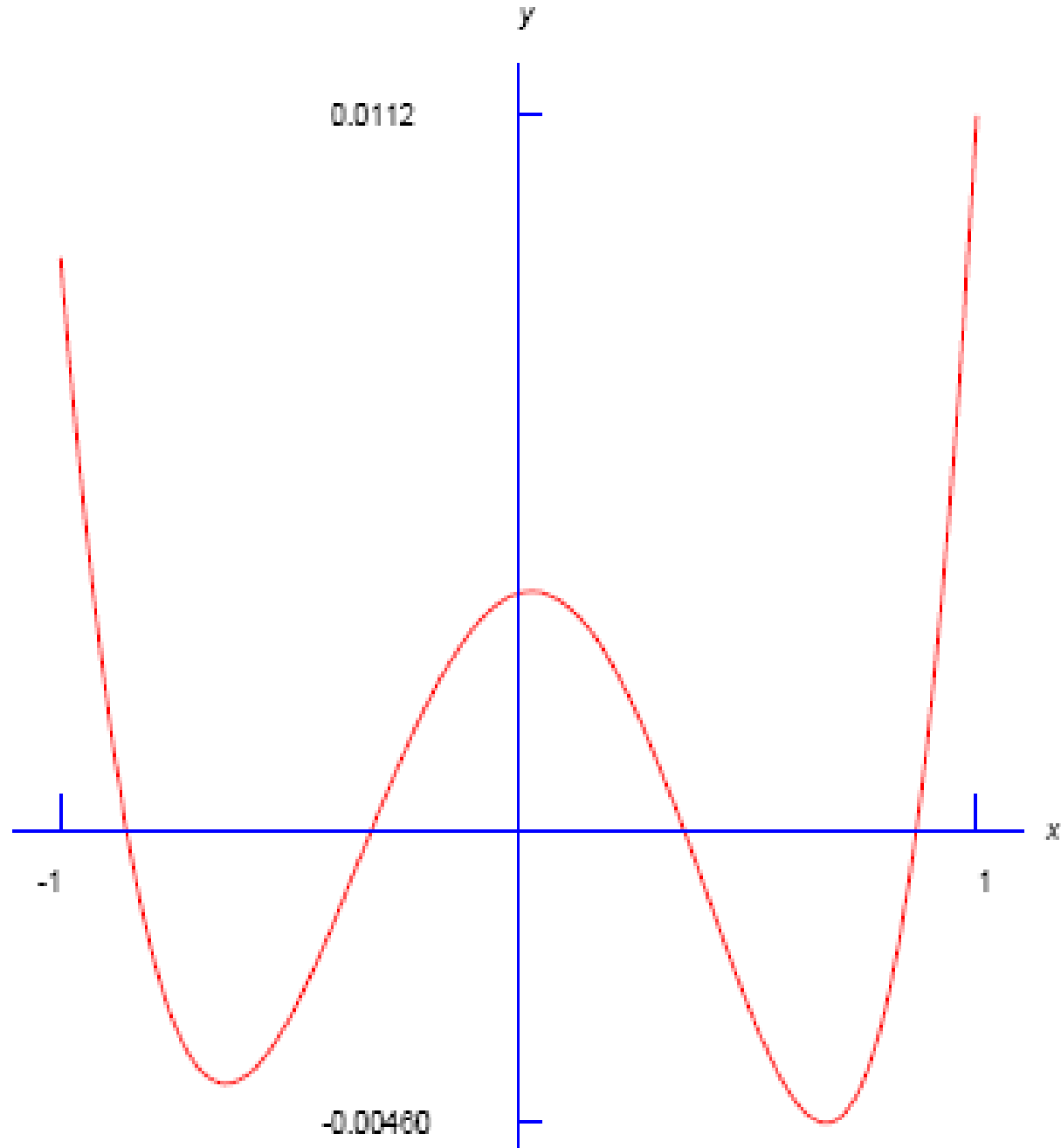
$$(f, g) = \int_a^b f(x)g(x) dx$$

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

$j$	0	1	2	3
$\beta_j$	2.35040	0.73576	0.14313	0.02013

$$l_3(x) = \sum_{j=0}^3 \beta_j P_j(x), \quad \beta_j = \frac{(f, P_j)}{(P_j, P_j)}$$

→  $l_3(x) = .996294 + .997955x + .536722x^2 + .176139x^3$



Error in the cubic least squares approximation to  $e^x$

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











