
Numerical Analysis and Computational Mathematics

Fall Semester 2019 - Section CSE

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Solutions - Approximation of functions and data

Solution I (MATLAB, tutorial)

Solutions are reported in the tutorial.

Solution II (MATLAB)

a) We execute the following commands:

```
f = @(x) sin( x );    a = 0;    b = 3 * pi;
n_vect = 1 : 7;      % vector containing all the degrees of desired polynomials
x_values = linspace( a, b, 1001 );
f_values = f( x_values );
for n = n_vect      % for all the degrees in n_vect
    x_nodes = linspace( a, b, n + 1 );
    y_nodes = f( x_nodes );
    P = polyfit( x_nodes, y_nodes, n );
    P_values = polyval( P, x_values );
    figure( n );
    plot( x_values, P_values, '-k', ...
          x_values, f_values, '—k', x_nodes, y_nodes, 'xk' );
    legend( '\Pi_n f(x)', 'f(x)', '(x_i, y_i)' );
end
```

We obtain the results reported in Figure 1 for example for $n = 2, 3, 5$, and 6. We observe the convergence of the interpolating polynomials $\Pi_n f(x)$ to $f(x)$ for increasing values of n . For $n = 3$ we observe that the couples $\{(x_i, f(x_i))\}$ for $i = 0, 1, 2, 3$ are aligned on a straight, horizontal line for which we obtain $\Pi_3 f(x) = c \in \mathbb{R}$, i.e. a polynomial of degree $n = 0$; specifically, we obtain that $\Pi_3 f(x) = 0$.

b) We insert the computation of the error in the MATLAB commands executed at point a), e.g.:

```
f = @(x) sin( x );    a = 0;    b = 3 * pi;
n_vect = 1 : 7;      % vector containing all the degrees of desired polynomials
x_values = linspace( a, b, 1001 );
```

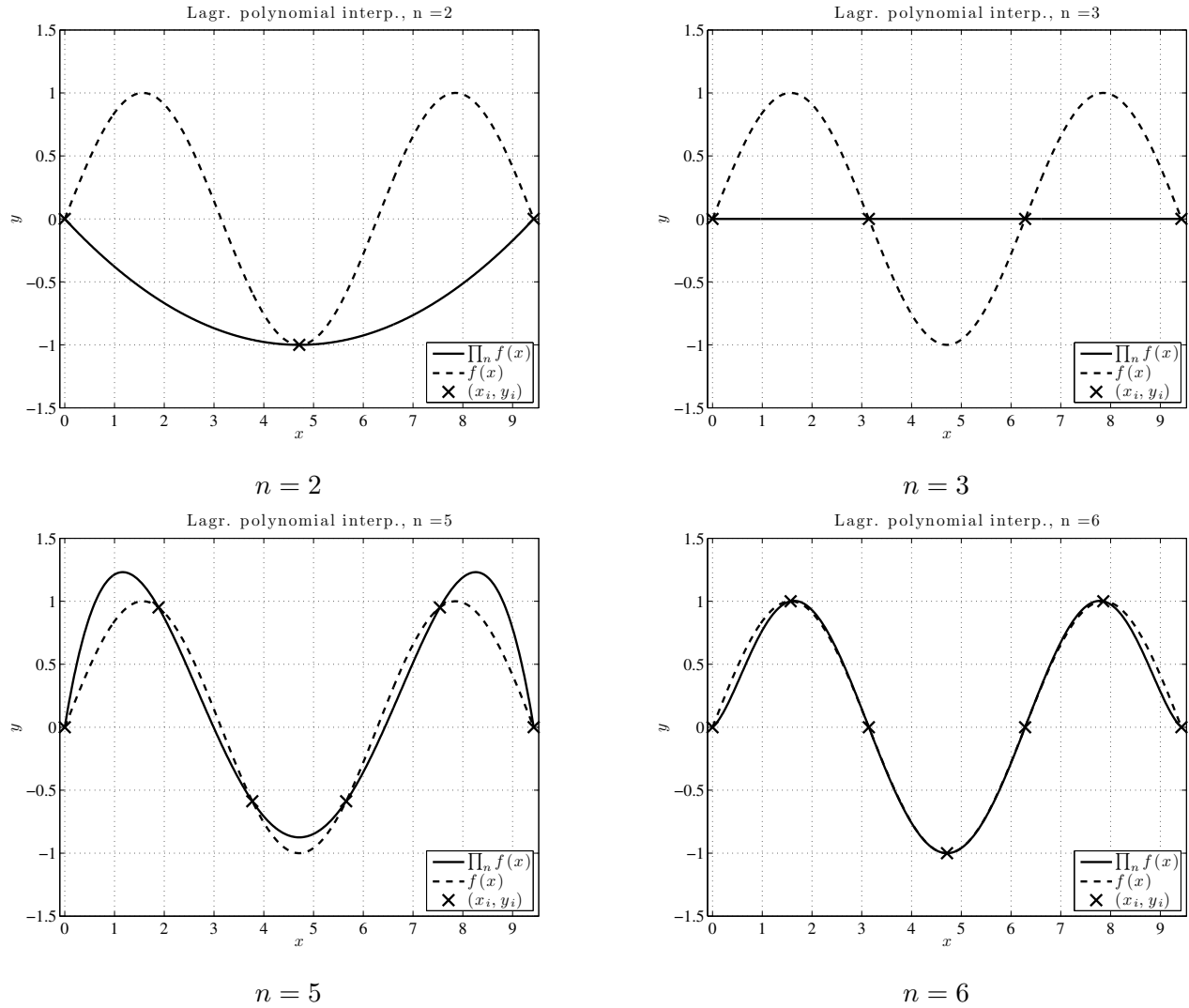


Figure 1: Interpolating polynomials $\Pi_n f(x)$ of the function $f(x) = \sin(x)$ at equally spaced nodes in $I = [0, 3\pi]$ for $n = 2, 3, 5$, and 6 .

```

f_values = f( x_values );
err = [ ]; % initialization of the vector containing the true errors
for n = n_vect
    x_nodes = linspace( a, b, n + 1 );
    y_nodes = f( x_nodes );
    P = polyfit( x_nodes, y_nodes, n );
    P_values = polyval( P, x_values );
    err = [ err, max( abs( P_values - f_values ) ) ]; % append errors to err
end
err
% err =
%      1.0000      1.5925      1.0000      0.6363      0.4228      0.1301      0.0895
plot( n_vect, err, '-ko' );

```

As we can observe from Figure 1(left) the error $e_n(f)$ is decreasing when n is increasing.

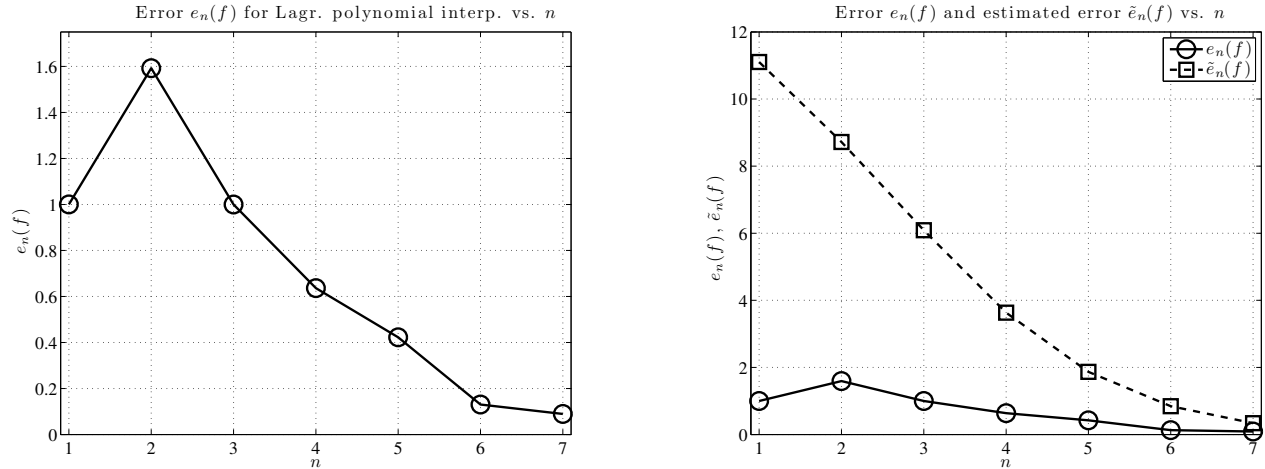


Figure 2: Errors $e_n(f)$ vs. n for the interpolating polynomials $\Pi_n f(x)$ of the function $f(x) = \sin(x)$ (left) and comparison with the error estimators $\tilde{e}_n(f)$ (right).

- c) We observe that $\max_{x \in I} |f^{(n+1)}(x)| = 1$, since $f^{(1)}(x) = \cos(x)$, $f^{(2)}(x) = -\sin(x)$, $f^{(3)}(x) = -\cos(x)$, ..., and $x \in I = [0, 3\pi]$. As consequence, the error estimator reads $\tilde{e}_n(f) = \frac{1}{4(n+1)} \left(\frac{b-a}{n}\right)^{n+1}$, which is monotonically decreasing when n increases. We plot in Figure 2(right) the error estimator $\tilde{e}_n(f)$ in comparison with the error $e_n(f)$ by means of the following commands:

```
err_estimated = [ ];
for n = n_vect
    df_max = 1; % for all n and x \in I=[0,3 *pi]
    err_estimated = [ err_estimated, ...
        1 / ( 4 * ( n + 1 ) ) * ( ( b - a ) / n ) ^ ( n + 1 ) * df_max ];
end
err_estimated
% err_estimated =
%    11.1033    8.7205    6.0881    3.6310    1.8689    0.8427    0.3375
plot( n_vect, err, '-ko', n_vect, err_estimated, '-ks' );
```

We verify that $e_n(f) \leq \tilde{e}_n(f)$ for all n . Since $\lim_{n \rightarrow \infty} \tilde{e}_n(f) = 0$ we have that $\lim_{n \rightarrow \infty} e_n(f) = 0$, i.e. the polynomial $\Pi_n f(x)$ converges to $f(x)$ for all $x \in I$ by increasing n .

Solution III (MATLAB)

- a) The Lagrange interpolating polynomial of degree n for $f(x)$ is $\Pi_n f(x) = \sum_{k=0}^n f(x_k) \varphi_k(x)$, where $\varphi_k(x) := \prod_{i=0, i \neq k}^n \frac{x-x_i}{x_k-x_i}$ are the Lagrange characteristic functions and x_i distinct nodes

with $i = 0, \dots, n$. For $n = 2$, we calculate $\varphi_k(x)$ for $k = 0, 1, 2$ as:

$$\begin{aligned}\varphi_0(x) &= \frac{x - x_1}{x_0 - x_1} \frac{x - x_2}{x_0 - x_2} = x^2 - \frac{5}{2}x + 1, \\ \varphi_1(x) &= \frac{x - x_0}{x_1 - x_0} \frac{x - x_2}{x_1 - x_2} = -\frac{4}{3}x^2 + \frac{8}{3}x, \\ \varphi_2(x) &= \frac{x - x_0}{x_2 - x_0} \frac{x - x_1}{x_2 - x_1} = \frac{1}{3}x^2 - \frac{1}{6}x.\end{aligned}$$

By observing that $f(x_0) = -2$, $f(x_1) = -\frac{11}{8}$, and $f(x_2) = 2$, we obtain $\Pi_2 f(x) = \frac{1}{2}x^2 + x - 2$.

- b) In this case we have $\varphi_0(x) = \frac{1}{2}x^2 - \frac{3}{2}x + 1$, $\varphi_1(x) = -x^2 + 2x$, and $\varphi_2(x) = \frac{1}{2}x^2 - \frac{1}{2}x$. By observing that $f(x_0) = -2$, $f(x_1) = 0$, and $f(x_2) = 2$, we obtain $\Pi_2 f(x) = 2x - 2$ which is a polynomial of degree 1. The result is due to the fact that the couples $\{(x_i, f(x_i))\}$ for $i = 0, \dots, n$ are aligned on a straight line.
- c) It is sufficient to observe that $f(x)$ is polynomial of degree 3 and therefore we have $\Pi_3 f(x) \equiv f(x)$.

Solution IV (MATLAB)

- a) We execute the following commands to compare the interpolating polynomials $\Pi_n f(x)$ with $f(x)$ in Figure 3.

```
f = @(x) 1 ./ ( 1 + x.^2 );    a = -5;    b = 5;
n_vect = [ 2 4 8 12 ];
x_values = linspace( a, b, 1001 );
f_values = f( x_values );
for n = n_vect
    x_nodes = linspace( a, b, n + 1 );
    y_nodes = f( x_nodes );
    P = polyfit( x_nodes, y_nodes, n );
    P_values = polyval( P, x_values );
    figure( n );
    plot( x_values, P_values, '-k', ...
          x_values, f_values, '-k', x_nodes, y_nodes, 'xk' );
    legend( '\Pi_n f(x)', 'f(x)', '(x_i, y_i)' );
end
```

We observe that oscillations of the polynomials $\Pi_n f(x)$ appear at the extrema of the interval I for n “large”, thus highlighting the so called Runge phenomenon; the amplitude of these oscillations generally increases with n .

- b) We plot the error $e_n(f)$ vs. n in Figure 4 by modifying the MATLAB commands at point a):

```
err = [ ];
for n = n_vect
    x_nodes = linspace( a, b, n + 1 );
    y_nodes = f( x_nodes );
    P = polyfit( x_nodes, y_nodes, n );
    P_values = polyval( P, x_values );
```

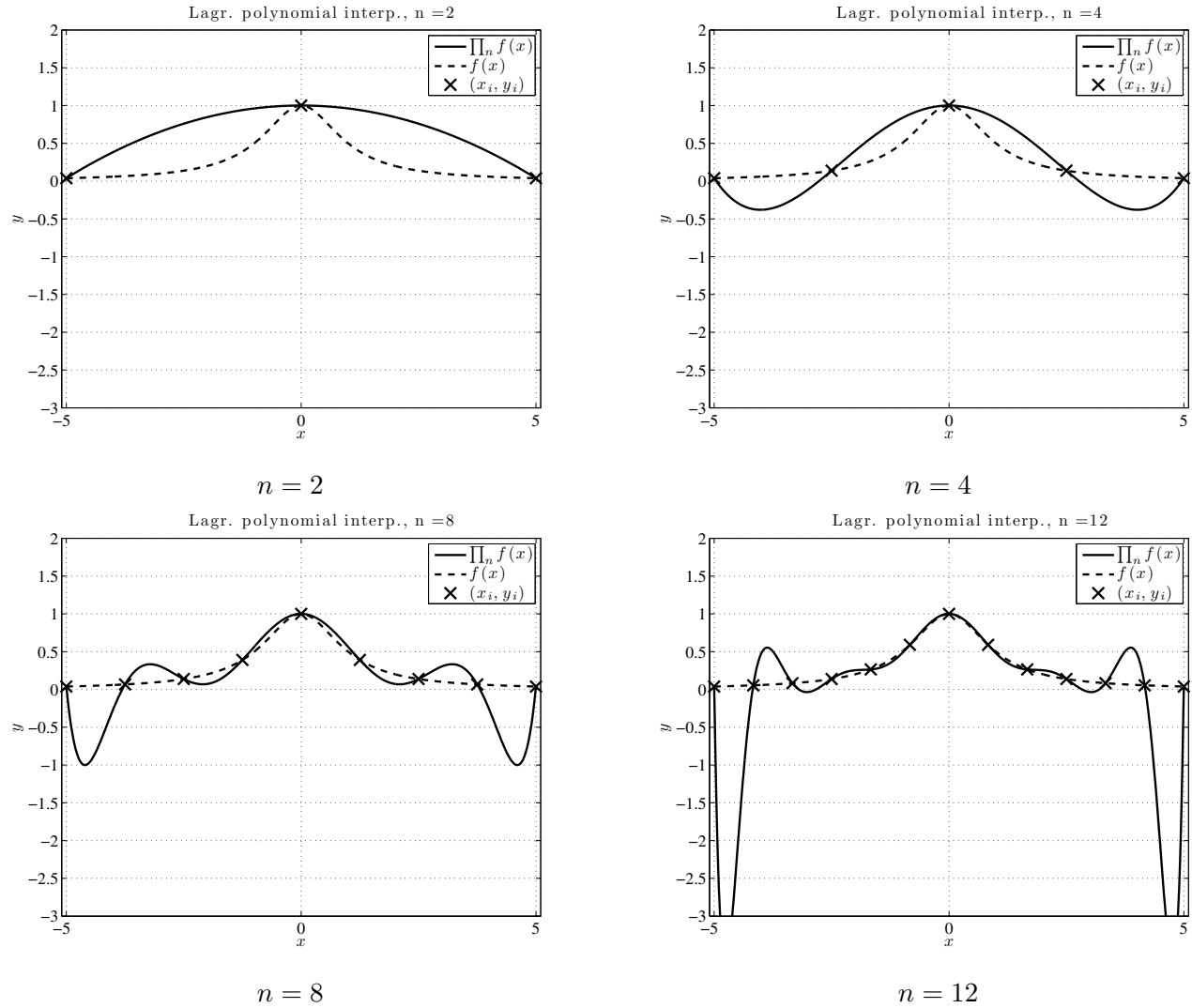


Figure 3: Interpolating polynomials $\Pi_n f(x)$ of the function $f(x) = \frac{1}{1+x^2}$ at equally spaced nodes in $I = [-5, 5]$ for $n = 2, 4, 8$, and 12 .

```

err = [ err, max( abs( P.values - f.values ) ) ];
end
err
% err =
%      0.6462      0.4384      1.0452      3.6630
figure; plot( n_vect, err, '-ko' );

```

Accordingly to the comments at point a), we see that the error $e_n(f)$ increases for n increasing due to the insurgency of the Runge phenomenon. The latter is a phenomenon which may occur when the value of $\max_{x \in I} |f^{(n+1)}(x)|$ considerably increases with n .

- c) We repeat point a) by using the Chebyshev–Gauss–Lobatto nodes in place of the equally spaced ones to determine the corresponding interpolating polynomials, which we denote as $\Pi_n^c f(x)$. In MATLAB, we use the following commands to obtain the results of Figure 5.

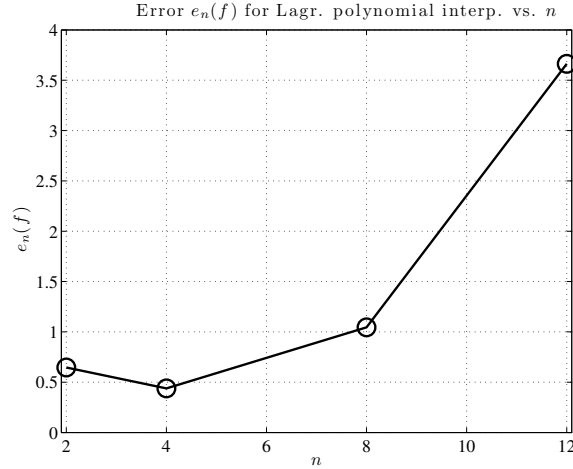


Figure 4: Errors $e_n(f)$ vs. n for interpolating polynomials $\Pi_n f(x)$ of the function $f(x) = \frac{1}{1+x^2}$ at equally spaced nodes in $I = [-5, 5]$; $n = 2, 4, 8$, and 12 .

```
for n = n_vect
    x_nodes_c = (a+b)/2 + (b-a)/2 * ( - cos( pi * [ 0 : n ] / n ) );
    y_nodes_c = f( x_nodes_c );
    P_c = polyfit( x_nodes_c, y_nodes_c, n );
    P_c_values = polyval( P_c, x_values );
    figure( n + 100);
    plot( x_values, P_c_values, '-k', ...
          x_values, f_values, '-k', x_nodes_c, y_nodes_c, 'xk' );
    legend( '$\prod_n f(x)$', '$f(x)$', '$(x_i, y_i)$' );
end
```

We observe that the interpolating polynomials $\Pi_n^c f(x)$ converge to $f(x)$ for increasing values of n . In Figure 6 we compare the interpolating polynomials $\Pi_8^c f(x)$ and $\Pi_8 f(x)$ with $f(x)$.

- d) By repeating point b) for the Chebyshev–Gauss–Lobatto nodes, we obtain that the error $e_n^c(f)$ associated to $\Pi_n^c f(x)$ decreases for increasing values of n (see Figure 7). We use the following MATLAB commands.

```
err_c = [ ];
for n = n_vect
    x_nodes_c = (a+b)/2 + (b-a)/2 * ( - cos( pi * [ 0 : n ] / n ) );
    y_nodes_c = f( x_nodes_c );
    P_c = polyfit( x_nodes_c, y_nodes_c, n );
    P_values_c = polyval( P_c, x_values );
    err_c = [ err_c, max( abs( P_values_c - f_values ) ) ];
end
err_c
% err_c =
%      6.4623e-01      4.5998e-01      2.0468e-01      8.4396e-02
plot( n_vect, err_c, '-ks' );
```

The result is justified by the fact that the use of the Chebyshev–Gauss–Lobatto nodes ensures that $\lim_{n \rightarrow \infty} e_n^c(f) = 0$ for $f(x) \in C^{(n+1)}(I)$.

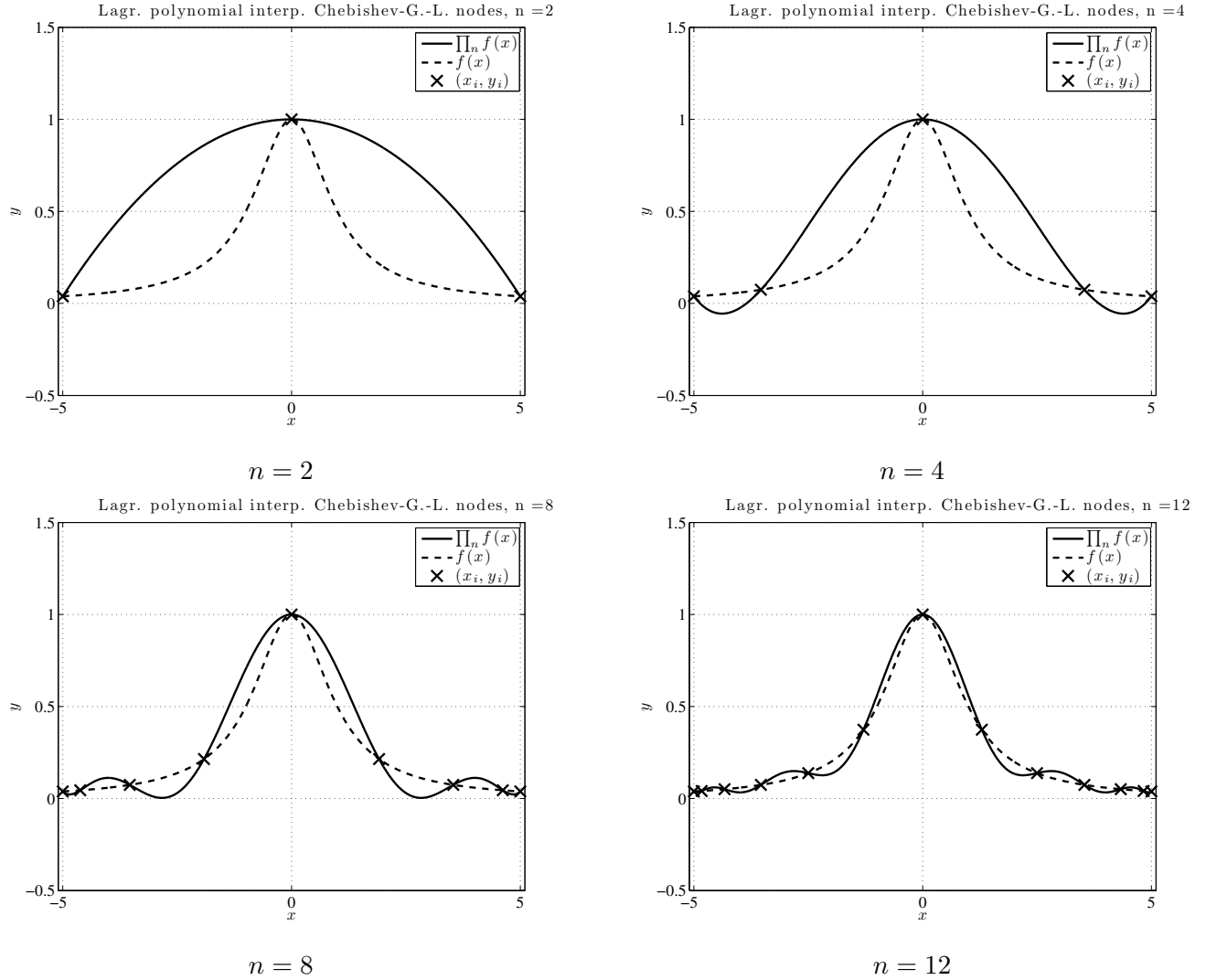


Figure 5: Interpolating polynomials $\Pi_n f(x)$ of the function $f(x) = \frac{1}{1+x^2}$ at the Chebyshev–Gauss–Lobatto nodes in $I = [-5, 5]$ for $n = 2, 4, 8$, and 12 .

Solution V (Theoretical)

- a) In general, given a function $f(x) \in C^{(n+1)}(I)$ with $I = [a, b]$ and the corresponding interpolating polynomial $\Pi_n f(x)$ of degree n defined at $n + 1$ equally spaced nodes $\{x_i\}_{i=0}^n$, we have the following estimate for the error $e_n(f) := \max_{x \in I} |f(x) - \Pi_n f(x)|$:

$$e_n(f) \leq \tilde{e}_n(f), \quad \text{where } \tilde{e}_n(f) := \frac{1}{4(n+1)} \left(\frac{b-a}{n} \right)^{n+1} \max_{x \in I} |f^{(n+1)}(x)|.$$

Specifically, for $f(x) = \sin\left(\frac{x}{3}\right)$, we obtain that $f^{(1)}(x) = \frac{1}{3} \cos\left(\frac{x}{3}\right)$, $f^{(2)}(x) = -\frac{1}{9} \sin\left(\frac{x}{3}\right)$, $f^{(3)}(x) = -\frac{1}{27} \cos\left(\frac{x}{3}\right)$, ...; as consequence, since $I = [a, b] = [0, 1]$, we deduce that $\max_{x \in I} |f^{(n+1)}(x)| \leq \frac{1}{3^{n+1}}$. By renaming $\tilde{e}_n(f)$ and using the previous result, we obtain that:

$$e_n(f) \leq \tilde{e}_n(f), \quad \text{with } \tilde{e}_n(f) = \frac{1}{4(n+1)(3n)^{n+1}}.$$

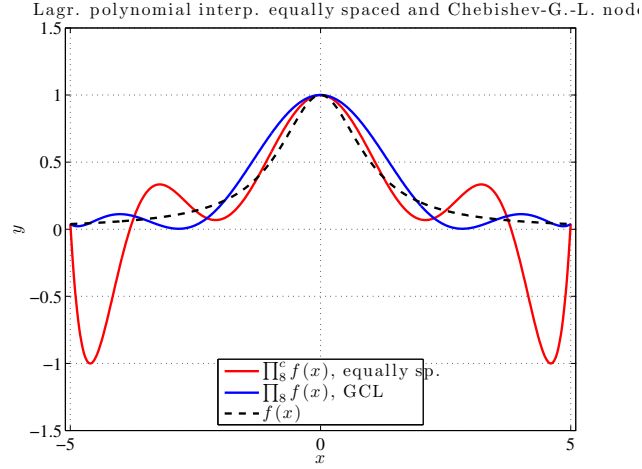


Figure 6: Interpolating polynomials $\Pi_8^c f(x)$ and $\Pi_8 f(x)$ of the function $f(x) = \frac{1}{1+x^2}$ at the Chebyshev–Gauss–Lobatto and equally spaced nodes in $I = [-5, 5]$, respectively.

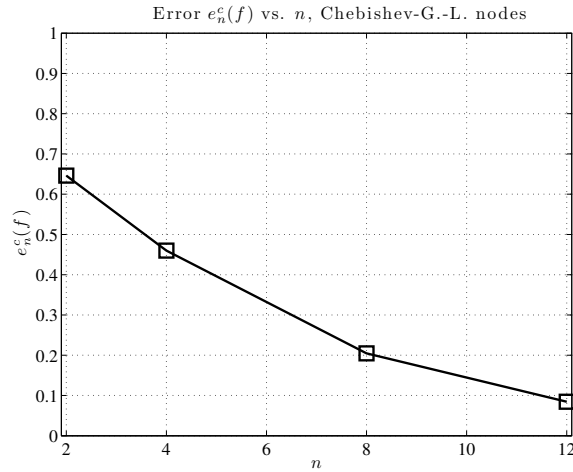


Figure 7: Errors $e_n^c(f)$ vs. n for interpolating polynomials $\Pi_n^c f(x)$ of the function $f(x) = \frac{1}{1+x^2}$ at the Chebyshev–Gauss–Lobatto nodes in $I = [-5, 5]$; $n = 2, 4, 8$, and 12 .

Since $\lim_{n \rightarrow \infty} \tilde{e}_n(f) = 0$, we have that the error $e_n(f)$ tends to zero for increasing values of n , i.e. $\lim_{n \rightarrow \infty} e_n(f) = 0$.

- b) We proceed by trial and error evaluating $\tilde{e}_n(f)$ for $n = 1, 2, 3, \dots$ until we guarantee that $e_n(f) \leq \tilde{e}_n(f) < 10^{-4}$. We obtain $\tilde{e}_1(f) = 1.3889 \cdot 10^{-2}$, $\tilde{e}_2(f) = 3.8580 \cdot 10^{-4}$, and, finally, $\tilde{e}_3(f) = 9.5260 \cdot 10^{-6}$. As consequence, the minimum number of equally spaced nodes in I necessary to ensure that $e_n(f) < 10^{-4}$ is $n + 1 = 4$.

- c) The Chebyshev–Gauss–Lobatto nodes in $I = [a, b]$ are determined by the formula:

$$x_i = \frac{a+b}{2} + \frac{b-a}{2} \hat{x}_i, \quad \text{where } \hat{x}_i := -\cos\left(\frac{\pi i}{n}\right), \quad \text{for } i = 0, \dots, n.$$

For $n = 3$, we have $\hat{x}_0 = -1$, $\hat{x}_1 = -\frac{1}{2}$, $\hat{x}_2 = \frac{1}{2}$, $\hat{x}_3 = 1$. Since $a = 0$ and $b = 1$, we obtain $x_0 = 0$, $x_1 = \frac{1}{4}$, $x_2 = \frac{3}{4}$, $x_3 = 1$.

- d) Since the Chebyshev–Gauss–Lobatto nodes are not equally spaced in I , we consider the following error estimate for the interpolating polynomials $\Pi_n f(x)$:

$$e_n(f) \leq \tilde{e}_n(f), \quad \text{where } \tilde{e}_n(f) := \frac{1}{(n+1)!} \max_{x \in I} |f^{(n+1)}(x)| \max_{x \in I} |\omega_n(x)|.$$

By setting $n = 3$, selecting $f(x) = \sin\left(\frac{x}{3}\right)$ in $I = [0, 1]$, observing that $\max_{x \in I} |\omega_3(x)| < 0.016$ (from Figure 1 of the exercises) and $\max_{x \in I} |f^{(4)}(x)| \leq \frac{1}{3^4}$ (from point a)), and renaming $\tilde{e}_3(f)$, we obtain that $e_3(f) \leq \tilde{e}_3(f)$, with $\tilde{e}_3(f) = 8.2305 \cdot 10^{-6}$. As consequence, the error $e_3(f)$ associated to the interpolating polynomial $\Pi_3 f(x)$ of the function $f(x)$ at the Chebyshev–Gauss–Lobatto nodes is inferior or equal to $8.2305 \cdot 10^{-6}$.