

MAT237 Numerical Analysis

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Solutions to midterm exam. Instructor: Ruben Agapito.

1 Roots of nonlinear equations

1. Use a plot to approximately locate all the roots of $f(x) = x^{-2} - \sin(x)$ in the interval $[1/2, 4\pi]$. Then find a pair of initial points for each root such that

$$q(x) = f(x_k) + \frac{f(x_k) - f(x_{k-1})}{x_k - x_{k-1}}(x - x_k).$$

converges to that root.

Solution. We want to find roots of the function f

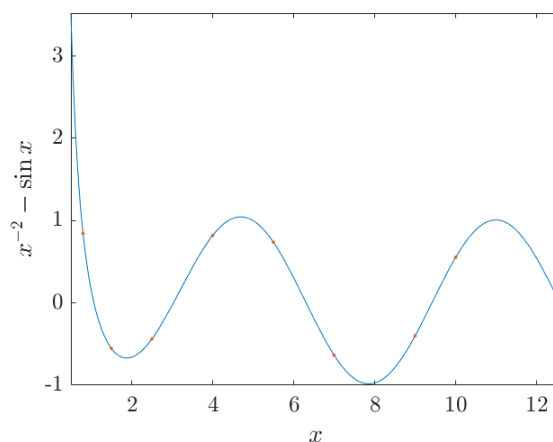


Figure 1: Plot of f in $[1/2, 4\pi]$

From the graph, it is clear that there are roots in the intervals $[0.8, 1.5]$, $[2.5, 4]$, $[5.5, 7]$ and $[9, 10]$. Using the secant method

```

1 function x = secant(f, x1, x2)
2 % Secant method for a scalar equation.
3 % Input:
4 %   f: objective function
5 %   x1, x2: initial root approximations
6 % Output
7 %   x: vector of root approximations (last is best)
8
9 % Operating parameters.
10 funtol = 100 * eps;
11 xtol = 100 * eps;
12 maxiter = 40;
13
14 x = [x1 x2];
15 dx = Inf;
16 y1 = f(x1);
17 k = 2;
18 y2 = 100;
19
20 while (abs(dx) > xtol) && (abs(y2) > funtol) && (k <
    maxiter)
21     y2 = f(x(k));
22     dx = -y2 * (x(k) - x(k - 1)) / (y2 - y1); % secant step
23     x(k + 1) = x(k) + dx;
24     k = k + 1;
25     y1 = y2; % current f-value becomes the old one next
        time
26 end
27
28 if k == maxiter, warning('Maximum number of iterations
    reached. '), end

```

the following code

```

1 clc;
2
3 set(0, 'defaulttextInterpreter', 'latex')
4 set(0, 'defaultAxesTickLabelInterpreter', 'latex');
5 set(0, 'defaultLegendInterpreter', 'latex');
6 set(0, 'defaultAxesFontSize', 15)
7 set(0, 'defaultLineLineWidth', 2.0);
8
9 f = @(x) x.^(-2) - sin(x);
10 interval = [1 / 2, 4 * pi];

```

```

11 fplot(f, interval);
12
13 x_axis = [0.8, 1.5, 2.5, 4, 5.5, 7, 9, 10];
14 y_axis = f(x_axis);
15 hold on, plot(x_axis, y_axis, '.');
16
17 for i = 1 : 4
18     root = secant(f, x_axis(2 * i - 1), x_axis(2 * i));
19     fprintf('root %d : %.10f\n', i, root(end));
20 end

```

outputs the approximated roots

```

root 1 : 1.0682235442
root 2 : 3.0326454184
root 3 : 6.3083168253
root 4 : 9.4134928032

```

□

2. Write a function that performs inverse quadratic interpolation for finding a root of f , give three initial estimates. To find the quadratic polynomial $q(y)$ passing through the three most recent points, use the built-in `polyfit` command. Test your function on the first problem from this section.

Solution. The following function performs inverse quadratic interpolation

```

1 function x = iqi(f, x1, x2, x3)
2     % Inverse quadratic interpolation method for a scalar
3     % equation.
4     % Input:
5     %   f: objective function
6     %   x1, x2, x3: initial estimates
7     % Output
8     %   x: vector of root approximations (last is best)
9
10    % Operating parameters.
11    funtol = 100 * eps;
12    xtol = 100 * eps;
13    maxiter = 40;
14
15    x = [x1 x2 x3];
16    y = f(x);
17    k = 0;
18    dx = abs(x3 - x2);
19
20    while (abs(dx) > xtol) && (abs(y(end)) > funtol) && (k
21        < maxiter)

```

```

20     coefficients = polyfit(y(k + 1 : k + 3), x(k + 1 :
        k + 3), 2);
21     x(k + 4) = polyval(coefficients, 0);
22     y(k + 4) = f(x(k + 4));
23     k = k + 1;
24     dx = abs(x(end) - x(end - 1));
25 end
26
27 if k == maxiter, warning('Maximum number of iterations
    reached. '), end

```

Examples 4.4.1 and 4.4.2 analyze the function $f(x) = xe^x - 2$ and Example 4.4.3 the function $f(x) = x + \cos(10x)$ with real domain. We are going to test our function with these examples

```

1  clc;
2  f = @(x) x.*exp(x) - 2;
3  x = [0.5 0.8 1];
4  estimates = iqi(f, x(1), x(2), x(3));
5  fprintf('Example 4.4.1 root : %.10f\n', estimates(end));
6
7  f = @(x) x + cos(10 * x);
8  x = [0.8 1.2 1];
9  estimates = iqi(f, x(1), x(2), x(3));
10 fprintf('Example 4.4.3 root : %.10f\n', estimates(end));

```

obtaining the following roots:

Example 4.4.1 root : 0.8526055020

Example 4.4.3 root : 0.9678884018

□

2 Polynomial interpolation

3. In each part, use

$$\Phi(x) = \prod_{i=0}^n (x - t_i)$$

to interpolate the given function using $n + 1$ evenly spaced nodes in the given interval. Plot each interpolant together with the exact function.

Solution. We are going to use the barycentric formula for the interpolating polynomial

```

1  function p = polyinterp(t, y)
2      % Polynomial interpolation by the barycentric formula.
3      % Input:

```

```

4 % t: interpolation nodes (vector, length n + 1)
5 % y: interpolation values (vector, length n + 1)
6 % Output:
7 % p: polynomial interpolant (function)
8
9 t = t(:); % column vector
10 n = length(t) - 1;
11 C = (t(end) - t(1)) / 4; % scaling factor to ensure
    stability
12 tc = t / C;
13
14 % Adding one node at a time, compute inverses of the
    weights.
15 omega = ones(n + 1, 1);
16 for m = 1 : n
17     d = (tc(1 : m) - tc(m + 1)); % vector of node
        differences
18     omega(1 : m) = omega(1 : m).*d; % update previous
19     omega(m + 1) = prod(-d); % compute the new one
20 end
21 w = 1./omega; % go from inverses to weights
22
23 p = @evaluate;
24
25 function f = evaluate(x)
26     % Compute interpolant, one value of x at a time.
27     f = zeros(size(x));
28     for j = 1:numel(x)
29         terms = w ./ (x(j) - t);
30         f(j) = sum(y.*terms) / sum(terms);
31     end
32
33     % Apply L'Hopital's Rule exactly.
34     for j = find(isnan(f(:)))' % divided by zero here
35         [~,idx] = min(abs(x(j) - t)); % node closest to x(
            j)
36         f(j) = y(idx); % value at node
37     end
38 end
39
40 end

```

□

b. $f(x) = \tanh x$, $n = 2, 3, 4$, $x \in [1, 2]$.

Solution. With the following code

```

1  clc;
2
3  set(0, 'defaulttextInterpreter', 'latex')
4  set(0, 'defaultAxesTickLabelInterpreter', 'latex');
5  set(0, 'defaultLegendInterpreter', 'latex');
6  set(0, 'defaultAxesFontSize', 15)
7  set(0, 'defaultLineLineWidth', 2.0);
8
9  f = @(x) tanh(x);
10 interval = [0.5 2];
11 fplot(f, interval), hold on
12
13 for n = 2 : 4
14     x = linspace(interval(1), interval(2), n + 1)';
15     p = polyinterp(x, f(x));
16     fplot(p, interval), hold on
17 end
18
19 hold off

```

we obtain

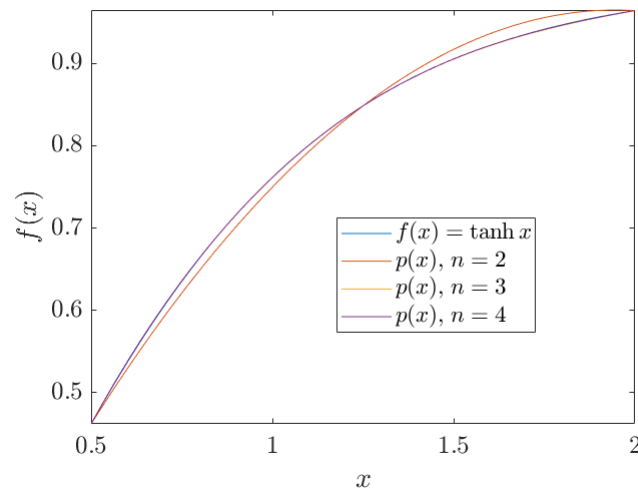


Figure 2: Polynomial interpolations of $\tanh x$ in $[0.5, 2]$

□

d. $f(x) = |x|$, $n = 2, 3, 4$, $x \in [-2, 1]$.

Solution. With the following code

```

1  clc;
2
3  set(0, 'defaulttextInterpreter', 'latex')
4  set(0, 'defaultAxesTickLabelInterpreter', 'latex');
5  set(0, 'defaultLegendInterpreter', 'latex');
6  set(0, 'defaultAxesFontSize', 15)
7  set(0, 'defaultLineLineWidth', 2.0);
8
9  f = @(x) abs(x);
10 interval = [-2 1];
11 fplot(f, interval), hold on
12
13 for n = 2 : 4
14     x = linspace(interval(1), interval(2), n + 1)';
15     p = polyinterp(x, f(x));
16     fplot(p, interval), hold on
17 end
18
19 hold off

```

we obtain

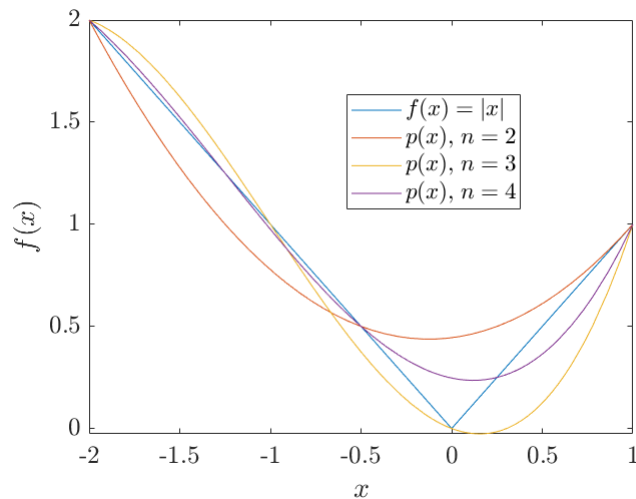


Figure 3: Polynomial interpolations of $|x|$ in $[-2, 1]$

□

4. Compute the barycentric weights numerically using $n + 1$ equally spaced nodes in $[-1, 1]$ for $n = 30, 60, 90$. On a single graph, plot their absolute values

as a function of t_i (which always ranges between -1 and 1) using a log-linear scale.

Solution. Modifying the barycentric interpolation function to return only the vector of weights

```

1 function w = barycentric_weights(t)
2     % Polynomial interpolation by the barycentric formula.
3     % Input:
4     %   t: interpolation nodes (vector, length n + 1)
5     % Output:
6     %   w: weights (vector, length n + 1)
7
8     t = t(:); % column vector
9     n = length(t) - 1;
10    C = (t(end) - t(1)) / 4; % scaling factor to ensure
        stability
11    tc = t / C;
12
13    % Adding one node at a time, compute inverses of the
        weights.
14    omega = ones(n + 1, 1);
15    for m = 1 : n
16        d = (tc(1 : m) - tc(m + 1)); % vector of node
            differences
17        omega(1 : m) = omega(1 : m).*d; % update previous
18        omega(m + 1) = prod(-d); % compute the new one
19    end
20    w = 1./omega; % go from inverses to weights
21 end

```

we plot the absolute values of the weights using a log-linear scale.

```

1 clc;
2
3 set(0, 'defaulttextInterpreter', 'latex')
4 set(0, 'defaultAxesTickLabelInterpreter', 'latex');
5 set(0, 'defaultLegendInterpreter', 'latex');
6 set(0, 'defaultAxesFontSize', 15)
7 set(0, 'defaultLineLineWidth', 2.0);
8
9 interval = [-1 1];
10
11 for n = 30:30:90
12     x = linspace(interval(1), interval(2), n + 1);
13     w = barycentric_weights(x);
14     semilogy(x, abs(w)), hold on
15 end

```

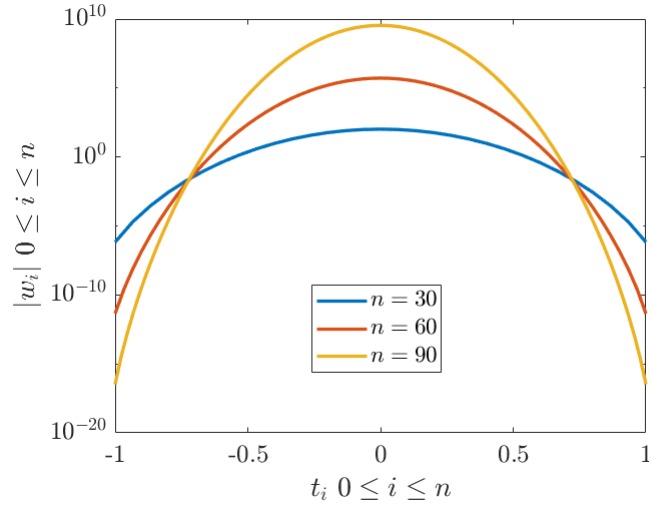



Figure 4: Absolute value of barycentric weights

□

5. For each case, compute the polynomial interpolant using n second-kind Chebyshev nodes in $[-1, 1]$ for $n = 4, 8, 12, \dots, 60$. At each value of n , compute the infinity-norm error (that is, $\max |p(x) - f(x)|$ evaluated for at least 4000 values of x). Using a log-linear scale, plot the error as a function of n , then determine a good approximation to the constant k in

$$\max_{x \in [-1, 1]} |f(x) - p(x)| \leq ck^{-n},$$

where f is analytic in an open real interval containing $[-1, 1]$, $c > 0$ and $k > 1$.

a. $f(x) = 1/(25x^2 + 1)$.

Solution. Computing the infinity-norm error of 4001 evenly spaced points between -1 and 1 of each polynomial interpolant, a good approximation for k is 1.85 using $c = 1$.

```

1  clc ;
2
3  set(0, 'defaulttextInterpreter', 'latex')
4  set(0, 'defaultAxesTickLabelInterpreter', 'latex');
5  set(0, 'defaultLegendInterpreter', 'latex');
6  set(0, 'defaultAxesFontSize', 15)
7  set(0, 'defaultLineLineWidth', 2.0);
8
9  f = @(x) 1./(25 * x.^2 + 16);

```

```

10
11 interval = [-1 1];
12 x = linspace(interval(1), interval(2), 4001);
13 nodes = linspace(4, 60, 4);
14 len = length(nodes);
15 ines = zeros(len);
16
17 for k = 1 : len
18     n = nodes(k);
19     p = chebinterp(f, n);
20     % Infinite norm error: max|p(x) - f(x)|
21     ine = 0;
22     for value = x
23         ine = max(ine, abs(p(value) - f(value)));
24     end
25     ines(k) = ine;
26 end
27
28 g = @(x) 1./((1.85).^x);
29 semilogy(nodes, ines), hold on
30 semilogy(nodes, g(nodes));

```

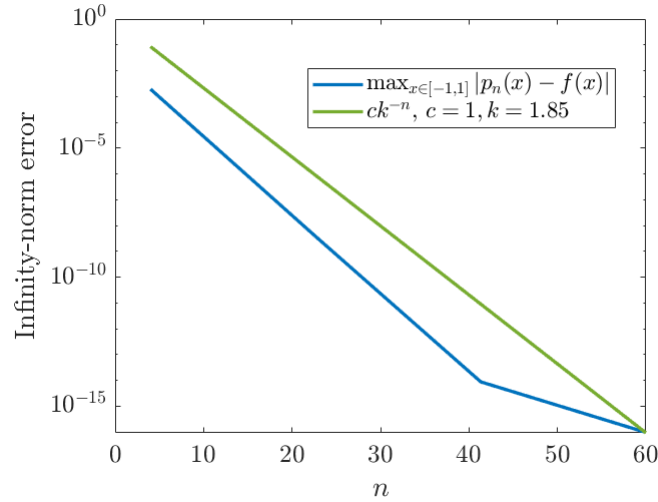


Figure 5: Infinity norm error of $f(x) = 1/(25x^2 + 1)$ and a bounding function

□

c. $f(x) = \cosh(\sin x)$.

Solution. Computing the infinity-norm error of 4001 evenly spaced points between -1 and 1 of each polynomial interpolant, a good approximation for k is 1.75 using $c = 1$.

```

1  clc ;
2
3  set(0, 'defaulttextInterpreter', 'latex')
4  set(0, 'defaultAxesTickLabelInterpreter', 'latex');
5  set(0, 'defaultLegendInterpreter', 'latex');
6  set(0, 'defaultAxesFontSize', 15)
7  set(0, 'defaultLineLineWidth', 2.0);
8
9  f = @(x) cosh(sin(x));
10
11 interval = [-1 1];
12 x = linspace(interval(1), interval(2), 4001);
13 nodes = linspace(4, 60, 4);
14 len = length(nodes);
15 ines = zeros(len);
16
17 for k = 1 : len
18     n = nodes(k);
19     p = chebinterp(f, n);
20     % Infinite norm error: max|p(x) - f(x)|
21     ine = 0;
22     for value = x
23         ine = max(ine, abs(p(value) - f(value)));
24     end
25     ines(k) = ine;
26 end
27
28 g = @(x) 1./((1.85).^x);
29 semilogy(nodes, ines), hold on
30 semilogy(nodes, g(nodes));

```

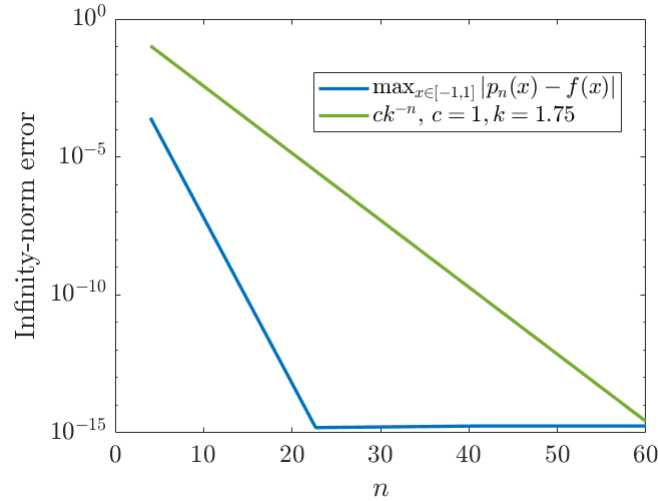


Figure 6: Infinite norm error of $f(x) = \cosh(\sin(x))$ and a bounding function

□

6. Write a function `function p = chebinterp(f, n)` that returns a function representing the polynomial interpolant of the function f using $n+1$ Chebyshev second-kind nodes (assuming $-1 \leq x \leq 1$). You should use

$$w_k = (-1)^k d_k, d_k = \begin{cases} 1/2 & \text{if } k = 0 \text{ or } k = n, \\ 1 & \text{otherwise} \end{cases}$$

to compute the barycentric weights directly, rather than using the method of finding ω and w in the barycentric formula. Test your function by revisiting the example with $f(x) = 1/(x^2 + 16)$ to use Chebyshev rather than equally spaced nodes.

Solution. The following function

```

1 function p = chebinterp(f, n)
2   % Polynomial interpolation by Chebyshev points of the
   % second kind.
3   % Input:
4   %   f: objective function (dom(f) = [-1, 1])
5   %   n: n + 1 Chebyshev second-kind nodes
6   % Output:
7   %   p: polynomial interpolant (function)
8   theta = linspace(0, pi, n + 1)';
9   nodes = -cos(theta);
10  p = polychebinterp(nodes, f(nodes));
11 end

```

uses the Chebyshev second-kind nodes to compute the polynomial interpolant with the barycentric formula computing the barycentric weights directly

```

1 function p = polychebinterp(t, y)
2     % Polynomial interpolation by the barycentric formula
3     % using Chebyshev points of the second kind.
4     % Input:
5     %   t: interpolation nodes (vector, length n + 1)
6     %   y: interpolation values (vector, length n + 1)
7     % Output:
8     %   p: polynomial interpolant (function)
9
10    t = t(:); % column vector
11    n = length(t) - 1;
12    C = (t(end) - t(1)) / 4; % scaling factor to ensure
    stability
13    tc = t / C;
14
15    % Barycentric weights using Chebyshev nodes
16    w = ones(n + 1, 1);
17    factor = -1;
18    for k = 1 : n + 1
19        if k == 1 || k == n + 1
20            d = 1 / 2;
21        else
22            d = 1;
23        end
24        w(k) = factor * d;
25        factor = factor * (-1);
26    end
27
28    p = @evaluate;
29
30    function f = evaluate(x)
31        % Compute interpolant, one value of x at a time.
32        f = zeros(size(x));
33        for j = 1:numel(x)
34            terms = w ./ (x(j) - t);
35            f(j) = sum(y.*terms) / sum(terms);
36        end
37
38        % Apply L'Hopital's Rule exactly.
39        for j = find(isnan(f(:)))' % divided by zero here
40            [~,idx] = min(abs(x(j) - t)); % node closest to x(
                j)
41            f(j) = y(idx); % value at node

```

```

42     end
43 end
44
45 end

```

In order to test our function, we are going to plot the polynomials p obtained using $n + 1$ nodes, for $n = 1, \dots, 5$. The decision of choosing small values for n was because from values greater than 4 the polynomials are almost f and we are not going to notice any differences in the plot.

```

1  clc;
2
3  set(0, 'defaulttextInterpreter', 'latex')
4  set(0, 'defaultAxesTickLabelInterpreter', 'latex');
5  set(0, 'defaultLegendInterpreter', 'latex');
6  set(0, 'defaultAxesFontSize', 15)
7  set(0, 'defaultLineLineWidth', 2.0);
8
9  f = @(x) 1./(x.^2 + 16);
10 interval = [-1 1];
11 fplot(f, interval), hold on
12 for n = 1 : 5
13     p = chebinterp(f, n);
14     fplot(p, interval), hold on
15 end
16 hold off

```

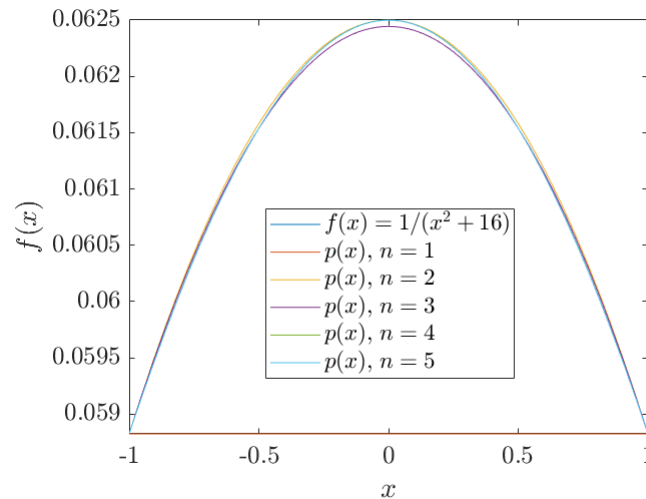


Figure 7: Polynomial interpolations of $1/(x^2 + 16)$ in $[-1, 1]$

□

7. Theorem 9.3.1 assumes that the function being approximated has infinitely many derivatives over $[-1, 1]$. But now consider the family of functions $f_m(x) = |x|^m$.

a. How many continuous derivatives over $[-1, 1]$ does f_m possess?

Solution. If m is even, we can write $m = 2k$ with k positive integer. Then

$$f_m(x) = |x|^{2k} = (|x|^2)^k = (x^2)^k = x^{2k} = x^m,$$

and because of being polynomial, f_m is infinitely differentiable and all its derivatives are continuous. When m is odd, we can write $m = 2k - 1$ with k positive integer. Then

$$f_m(x) = |x|^{2k-1} = |x| |x|^{2k-2} = |x| x^{2k-2} = |x| x^{m-1}.$$

Theorem 1. Given $f : [-1, 1] \rightarrow \mathbb{R}$ with $f(x) = c_n |x|^n$, where c is a constant positive number, has exactly n continuous derivatives.

Prueba. Let's prove it by induction. When $n = 1$, the derivative in $(0, 1]$ is $2cx$ and the derivative in $[-1, 0)$ is $-2cx$. By definition, at $x = 0$ we have

$$\lim_{h \rightarrow 0} \frac{|h|h - 0}{h} = \lim_{h \rightarrow 0} |h| = 0.$$

so the derivative is $2c|x|$ which is continuous and it doesn't have more derivatives because $|x|$ is not differentiable at $x = 0$. Suppose that $c_n |x|^n$ has exactly n continuous derivatives. Analyzing $c_{n+1} |x|^{n+1}$, in $(0, 1]$ we have $(n+2)c_{n+1}x^{n+1}$, in $[-1, 0)$ we have $-(n+2)c_{n+1}x^{n+1}$. By definition, at $x = 0$ we have

$$\lim_{h \rightarrow 0} \frac{c_{n+1}|h|h^{n+1}}{h} = \lim_{h \rightarrow 0} \frac{c_{n+1}|h|h}{h} = 0.$$

Then, the derivative is $(n+1)c_{n+1}|x|x^n = c_n |x|^n$. This function is continuous in our interval because is a product of continuous functions. Also, because of the inductive hypothesis, this function has n continuous derivative, so in total we have $n+1$ continuous derivatives. □

Using the theorem, we can conclude that f_m has $m-1$ continuous derivatives when m is odd. □

b. Compute the polynomial interpolant using n second-kind Chebyshev nodes in $[-1, 1]$ for $n = 10, 20, 30, \dots, 100$. At each value of n , compute the infinity-norm error. (that is, $\max |p(x) - f_m(x)|$ evaluated for at least 41000 values of x). Using a single log-log graph, plot the error as a function of n for all six values $m = 1, 3, 5, 7, 9, 11$.

Solution. We create a function called `abspow` defined as $\text{abspow}(x, m) = |x|^m$ to fix the exponent on each iteration and then use it as a one variable function of x in the interpolation.

```
1 function y = abspow(x, m)
2     y = abs(x).^m;
3 end
```

Then, for every odd value of m between 1 and 11, we are going to plot a log-log graph of the infinity-norm errors.

```
1 clc;
2
3 set(0, 'defaulttextInterpreter', 'latex')
4 set(0, 'defaultAxesTickLabelInterpreter', 'latex');
5 set(0, 'defaultLegendInterpreter', 'latex');
6 set(0, 'defaultAxesFontSize', 15)
7 set(0, 'defaultLineLineWidth', 2.0);
8
9 interval = [-1 1];
10 x = linspace(interval(1), interval(2), 41001);
11 nodes = 10 : 10 : 100;
12 len = length(nodes);
13
14 for m = 1 : 2 : 11
15     % Fix f(x) = |x|^m
16     f = @(x) abspow(x, m);
17     ines = zeros(len);
18     % For each n
19     for k = 1 : len
20         n = nodes(k);
21         p = chebinterp(f, n);
22         ine = 0;
23         % Find max |p(x) - f(x)|
24         for value = x
25             ine = max(ine, abs(p(value) - f(value)));
26         end
27         ines(k) = ine;
28     end
29     loglog(nodes, ines), hold on
30 end
31
32 hold off
```

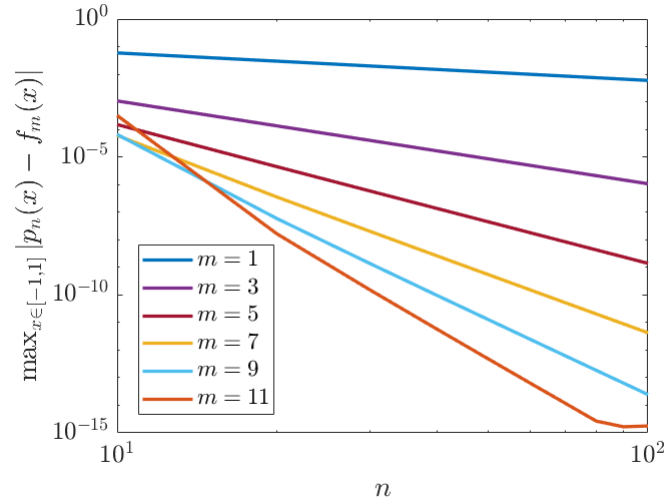



Figure 8: Infinity-norm error of $f(x) = |x|^m$ in $[-1, 1]$

□

c. Based on the result of parts (a) and (b), form a hypothesis about the asymptotic behaviour of the error for fixed m as $n \rightarrow \infty$.

Solution. From the graph in part (b), for a fixed value of m , we can form a hypothesis that

$$\ln \max_{x \in [-1, 1]} |f(x) - p(x)| = -\alpha \ln n + \beta$$

with $\alpha > 0$ and $0 < \beta < 1$. The infinity-norm error of f is

$$\begin{aligned} \max_{x \in [-1, 1]} |f(x) - p(x)| &= \exp(-\alpha \ln n + \beta) \\ &= \exp(-\alpha \ln n) \exp(\beta) \\ &= \exp(\beta) n^{-\alpha}. \end{aligned}$$

The theorem in the book requires an analytic function, however, even though our function only has $m - 1$ derivatives, experimentally the method converges and the infinity-norm error is $\mathcal{O}(n^{-\alpha})$. □