CSE 840: Computational Foundations of Artificial Intelligence March 10, 2025 Lecture 12: Power Series, Taylor Series Instructor: Vishnu Boddeti Scribe: Minh Nguyen

## **1** Power Series

**Definition 1** A series of the form

$$p(x) = \sum_{n=0}^{\infty} a_n x^n$$

is called a power series.

An infinite sum is called a **series**, and it consists of terms that are **powers** of x.

Theorem 2 (Radius of Convergence) For every power series

$$p(x) = \sum_{n=0}^{\infty} a_n x^n$$

there exists a constant  $r, 0 \le r \le \infty$ , called the radius of convergence such that:

- The series converges **absolutely** for all x with |x| < r (means that  $\sum_{n=0}^{\infty} a_n |x|^n$  converges, the sequence of partial sums  $P_N(x) := \sum_{n=0}^{N} a_n |x|^n$  converges "in the usual sense" as  $N \to \infty$ ).
- **Unclear** what happens when |x| = r.
- If |x| < r, the series even converges uniformly.

The radius of convergence only depends on the sequence  $(a_n)_n$  and can be computed by various formulae (if it exists):

- $\gamma = \frac{1}{L}$  where  $L = \limsup_{n \to \infty} (|a_n|)^{1/n}$
- $\gamma = \lim_{n \to \infty} \left| \frac{a_n}{a_{n+1}} \right|$

Example 3 Consider the power series:

$$p(x) = \sum_{n=0}^{\infty} n^c x^n$$
 for some constant c.

The radius of convergence is given by:

$$\gamma = \lim_{n \to \infty} \left| \frac{a_n}{a_{n+1}} \right| = \lim_{n \to \infty} \frac{n^c}{(n+1)^c} = \lim_{n \to \infty} \left( \frac{n}{n+1} \right)^c = 1.$$

*Case 1* <u>For c = -1:</u>

$$\sum_{n=1}^{\infty} \frac{1}{n} x^n,$$

the radius of convergence is r = 1.

• For x = +1, the series diverges:

$$\sum_{n=1}^{\infty} \frac{1}{n} \cdot x^n = \sum_{n=1}^{\infty} \frac{1}{n} \cdot 1^n = \sum_{n=1}^{\infty} \frac{1}{n} \to \infty.$$

- For x = -1, the series converges.
- For x > 1, the series diverges.

*Case 2* <u>For c = 0:</u>

$$\sum_{n=0}^{\infty} n^c x^n = \sum_{n=0}^{\infty} x^n,$$

the series diverges for |x| = r (both x = 1 and x = -1).

**Example 4** Exponential Series:

$$\exp(x) = \sum_{n=0}^{\infty} \frac{x^n}{n!}$$

has a radius of convergence  $r = \infty$ , since

$$\lim_{n \to \infty} \left| \frac{a_n}{a_{n+1}} \right| = \lim_{n \to \infty} \frac{(n+1)!}{n!} = n+1 \to \infty.$$

Example 5 The series:

$$\sum_{n=0}^{\infty} n! \, x^n$$

has a radius of convergence r = 0, since

$$\lim_{n \to \infty} \left| \frac{a_n}{a_{n+1}} \right| = \lim_{n \to \infty} \frac{n!}{(n+1)!} = \lim_{n \to \infty} \frac{1}{n+1} \to 0.$$

# From Power Series to Taylor Series

**Observation 6** Given a power series:

$$f(x_0+h) = \sum_{n=0}^{\infty} a_n h^n,$$

its derivative is:

$$f'(x_0+h) = (a_0 + a_1h + a_2h^2 + \dots)' = a_1 + 2a_2h + 3a_3h^2 + \dots = \sum_{n=1}^{\infty} n \cdot a_nh^{n-1}$$

For higher-order derivatives of the power series:

$$f^{(k)}(x_0+h) = \sum_{n=k}^{\infty} a_n \big( n \cdot (n-1) \cdot (n-2) \cdots (n-k+1) \big) h^{n-k}.$$

In particular, we have:

$$f^{(k)}(x_0) = a_k \cdot k! \quad \Rightarrow \quad \left| a_k = \frac{f^{(k)}(x_0)}{k!} \right|$$

**Theorem 7** Let  $f(x_0 + h) = \sum_{n=0}^{\infty} a_n h^n$  with r > 0. Then for h with |h| < r, we have:

$$f(x_0 + h) = \sum_{n=0}^{\infty} \frac{f^{(n)}(x_0)}{n!} h^n.$$

**Remark 8** Intuition: Start with a power series that converges. Then we have a nice formula that expresses the coefficients in terms of the derivatives of the function.

## Question

Does the theorem hold the other way around? That is, given any function (possibly with nice assumptions), can we simply build the series:

$$\sum_{n=0}^{\infty} \frac{f^{(n)}(x_0)}{n!} h^n$$

and "hope" that it converges to the function f(x)?

# 2 Taylor Series

## Intuition of Taylor's Theorem



- Linear Approximation:  $f(x_0 + h) = f(x_0) + f'(x_0) \cdot h + r(h) \cdot h$ , where  $x = x_0 + h$  and  $r(h) \xrightarrow{h \to 0} 0$ .
- <u>Quadratic Approximation:</u>  $f(x_0 + h) = f(x_0) + f'(x_0) \cdot h + \frac{1}{2}f''(x_0) \cdot h^2 + r(h) \cdot h^2$ , where  $r(h) \xrightarrow{h \to 0} 0$ .

**Theorem 9** Let  $I \subset \mathbb{R}$  be an open interval, and  $f : I \to \mathbb{R}$ . Suppose  $f \in C^{n+1}([a,b])$  and  $x_0 \in I$ . Define:

$$\begin{split} T_n(x_0,h) &:= \sum_{k=0}^n \frac{f^{(k)}(x_0)}{k!} \cdot h^k \quad (Taylor \ series \ up \ to \ degree \ n). \\ R_n(x_0,h) &:= \int_{x_0}^{x_0+h} \frac{(x+h-t)^n}{n!} f^{(n+1)}(t) \ dt \quad (Remainder \ term). \end{split}$$

Then:

$$f(x_0 + h) = T_n(x_0, h) + R_n(x_0, h).$$

**Proof Sketch:** The proof follows from the fundamental theorem of calculus and proceeds by induction on n.

• Base case (n = 0): We need to prove:

$$f(x_0 + h) = f(x_0) + \int_{x_0}^{x_0 + h} f'(t) dt.$$

This is equivalent to the fundamental theorem of calculus:

$$\int_{a}^{b} F'(x) \, dx = F(b) - F(a).$$

• Inductive step  $(n \rightarrow n+1)$ : Consider the function:

$$F(x_0 + h) = \frac{(x_0 + h - t)^{n+1}}{(n+1)!} f^{(n+1)}(t).$$

The proof proceeds by integrating and applying the fundamental theorem of calculus.

- Take its derivative.
- Integrate and exploit fundamental theorem.

 $\triangle$ 

**Theorem 10 (Taylor's Theorem with Lagrange Remainder)** Let  $I \subset \mathbb{R}$  be an interval,  $f : I \to \mathbb{R}$ , and  $f \in C^{n+1}(I)$ ,  $x_0 \in I$ . If  $h \in \mathbb{R}$  such that  $x_0 + h \in I$ . Then:

$$f(x_0 + h) = \sum_{k=0}^{n} \frac{f^{(k)}(x_0)}{k!} \cdot h^k + R_n(h),$$

where the first part is **n-th order Taylor polynomial**, and  $R_n(h)$  is the remainder term of order n. There exists  $\xi$  with  $\xi \in (x_0, x_0 + h)$  (or  $\xi \in (x_0 + h, x_0)$ ) such that:

$$R_n(h) = \frac{f^{(n+1)}(\xi)}{(n+1)!} \cdot h^{n+1}.$$

Often, this is written as:

$$f(x_0 + h) = \sum_{k=0}^{n} \frac{f^{(k)}(x_0)}{k!} \cdot h^k + O(h^{n+1}),$$

or, with  $x = x_0 + h$ :

$$f(x) = \sum_{k=0}^{n} \frac{f^{(k)}(x_0)}{k!} (x - x_0)^k + O((x - x_0)^{n+1}).$$

### **Proof of Theorem 10:**

$$F_{n,h}(t) := \sum_{k=0}^{n} \frac{f^{(k)}(t)}{k!} (x_0 + h - t)^k,$$

Note:

$$F_{n,h}(x_0) = T_n(x_0, h), \quad F_{n,h}(x_0 + h) = f(x_0 + h)$$

Define:

$$g_{n,h}(t) := (x_0 + h - t)^{n+1}, \quad g'_{n,h}(t) = -(n+1)(x_0 + h - t)^n$$

#### Generalized Mean Value Theorem

Using the generalized mean value theorem, we have:

$$\frac{F_{n,h}(x_0+h) - F_{n,h}(x_0)}{g_{n,h}(x_0+h) - g_{n,h}(x_0)} = \frac{F'_{n,h}(\xi)}{g'_{n,h}(\xi)},$$

where  $\xi \in (x_0, x_0 + h)$ .

Replacing  $F_{n,h}(x_0 + h)$  with  $f(x_0 + h)$ , we have:

$$f(x_0 + h) - T_n(x_0, h) = (g_{n,h}(x_0 + h) - g_{n,h}(x_0)) \cdot \frac{F'_{n,h}(\xi)}{g'_{n,h}(\xi)}$$
$$= \frac{h^{n+1} \cdot F'_{n,h}(\xi)}{(n+1) \cdot (h+x_0 - \xi)^n}$$

Left clause:

$$f(x_0 + h) - T_n(x_0, h) = F'_{n,h}(t)$$
  
=  $\frac{d}{dt} \sum_{k=0}^n \frac{f^{(k)}(t)}{k!} (h + x_0 - t)^k$   
=  $\sum_{k=0}^n \frac{f^{(k+1)}(t)}{k!} (h + x_0 - t)^k - \sum_{k=1}^n \frac{f^{(k)}(t)}{(k-1)!} (h + x_0 - t)^{k-1}$   
=  $\frac{f^{(n+1)}(t)}{n!} (h + x_0 - t)^n$ 

Substituting to right clause:

$$f(x_0+h) - T_n(x_0,h) = \frac{h^{n+1} \cdot \frac{f^{(n+1)}(\xi_p)}{n!} \cdot (h+x_0-\xi)^n}{(n+1) \cdot (h+x_0-\xi)^n}$$
$$= \frac{h^{n+1} \cdot f^{(n+1)}(\xi)}{(n+1)!}$$

Result:

$$f(x_0 + h) = T_n(x_0, h) + \frac{h^{n+1} \cdot f^{(n+1)}(\xi)}{(n+1)!}$$

г	_	_	_	
L				
L				
L				

**Theorem 11** Let  $f \in C^{\infty}(I)$ ,  $x_0 \in I$ , and  $h \in \mathbb{R}$  such that  $x_0 + h \in I$ . Define

$$T(x_0,h) := \lim_{n \to \infty} T_n(x_0,h) = \sum_{n=0}^{\infty} \frac{f^{(n)}(x_0)}{n!} \cdot h^n.$$

Then we have f(x) = T(x) if  $R_n(x_0, h) \xrightarrow{n \to \infty} 0$ .

For example, this is the case if there exist constants  $\alpha, c > 0$  such that

$$\left|\frac{f^{(n)}(t)}{f(t)}\right| \le \alpha \cdot c^n, \quad \forall t \in I, \quad \forall n \in \mathbb{N}.$$

This follows directly from the Lagrangian remainder.

#### Examples

• Exponential series:

$$\exp(x) = \sum_{n=0}^{\infty} \frac{x^n}{n!}$$
 (Power series with  $r = \infty$ )

The exponential function always coincides with its Taylor series.

Other examples include sin, cos, polynomials, and power series (analytic functions).

•  $f(x) = \log(x+1)$ , Taylor series around zero.

To prove: The convergence radius for the Taylor series is r = 1. For x outside of (-1, 1), the Taylor series does not make sense at all.

$$f(x) = \begin{cases} \exp\left(-\frac{1}{x^2}\right) & \text{if } x \neq 0, \\ 0 & \text{if } x = 0. \end{cases}$$

This function has the peculiar property that for all  $n \in \mathbb{N}$ :

$$f^{(n)}(0) = 0.$$

Consider the Taylor series about  $x_0 = 0$ . All terms will be zero, i.e.,  $\forall n : T_n(0, h) = 0$  and  $r = \infty$ .

$$f(x_0 + h) = T_n(x_0, h) + R_n(x_0, h)$$

$$T_n(x_0 = 0, h) = 0$$
 but  $f(0+h) = \exp\left(-\frac{1}{h^2}\right)$ 

Taylor series around  $x_0 = 0$  is zero. Function value around  $x_0 = 0$  is not zero.

$$\forall (x_0 + h) \neq 0, \quad T_n(x_0, h) \neq f(x_0 + h)$$

## 2.1 Taylor Series in Deep Learning

Taylor series can be used to approximate activation functions in deep learning.

### • Softmax Activation:

$$\operatorname{softmax}(z_i) = \frac{e^{z_i}}{\sum_{j=1}^{K} e^{z_j}}$$

Taylor expansions can approximate the exponential terms for efficient computation. The power series for  $e^x$  is:

$$e^x = \sum_{n=0}^{\infty} \frac{x^n}{n!} = 1 + x + \frac{x^2}{2!} + \frac{x^3}{3!} + \cdots$$

By substituting the power series expansion of  $e^{z_i}$  and  $e^{z_j}$ , we can approximate the softmax function:

softmax
$$(z_i) \approx \frac{1 + z_i + \frac{z_i^2}{2!} + \frac{z_i^3}{3!} + \cdots}{\sum_{j=1}^{K} \left(1 + z_j + \frac{z_j^2}{2!} + \frac{z_j^3}{3!} + \cdots\right)}$$

• Sigmoid Activation:

$$\sigma(x) = \frac{1}{1 + e^{-x}} \approx \frac{1}{2} + \frac{x}{4} - \frac{x^3}{48} + \cdots$$

• Tanh Activation:

$$\tanh(x) = \frac{e^x - e^{-x}}{e^x + e^{-x}} \approx x - \frac{x^3}{3} + \frac{2x^5}{15} - \cdots$$

• ReLU Activation:

$$\operatorname{ReLU}(x) = \max(0, x)$$

(No Taylor expansion needed, but approximations may be used for variations like Leaky ReLU.)

These approximations simplify calculations, especially in resource-constrained environments.