

## Lecture A1. Math Review

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- 1 I. Differentiation and integration
- 2 II. Numerical methods for finding a root
- 3 III. Matrix algebra
- 4 IV. Series and others

# I. Differentiation and integration

# Differentiation

## Definition 1 (differentiation)

**Differentiation** is the action of computing a derivative.

## Definition 2 (derivative)

The **derivative** of a function  $y = f(x)$  of a variable  $x$  is a measure of the rate at which the value  $y$  of the function changes with respect to (wrt., hereafter) the change of the variable  $x$ . It is notated as  $f'(x)$  and called **derivative** of  $f$  wrt.  $x$ .

## Remark 1

If  $x$  and  $y$  are real numbers, and if the graph of  $f$  is plotted against  $x$ , the derivative is the **slope of this graph** at each point.

### Definition 3 (differentiable)

If  $\lim_{h \rightarrow 0} \frac{f(x+h/2)-f(x-h/2)}{h}$  exists for a function  $f$  at  $x$ , we say the function  $f$  is **differentiable at  $x$** . That is,  $f'(x) = \lim_{h \rightarrow 0} \frac{f(x+h/2)-f(x-h/2)}{h}$ . If  $f$  is differentiable for all  $x$ , then we say  $f$  is **differentiable (everywhere)**.

### Remark 2

The followings are popular derivatives.

- $f(x) = x^p \Rightarrow f'(x) = px^{p-1}$  (polyomial)
- $f(x) = e^x \Rightarrow f'(x) = e^x$  (exponential)
- $f(x) = \log(x) \Rightarrow f'(x) = 1/x$  (log function; not differentiable at  $x = 0$ )

### Theorem 1

*Differentiation is linear. That is,  $h(x) = f(x) + g(x)$  implies  $h'(x) = f'(x) + g'(x)$ .*

## Theorem 2 (differentiation of product)

If  $h(x) = f(x)g(x)$ , then  $h'(x) = f'(x)g(x) + f(x)g'(x)$ .

### Exercise 1

Suppose  $f(x) = xe^x$ , find  $f'(x)$ .

## Theorem 3 (differentiation of fraction)

If  $h(x) = \frac{f(x)}{g(x)}$ , then  $h'(x) = \frac{f'(x)g(x) - f(x)g'(x)}{(g(x))^2}$ .

## Theorem 4 (composite function)

*If  $h(x) = f(g(x))$ , then  $h'(x) = f'(g(x)) \cdot g'(x)$ .*

## Exercise 2

*Suppose  $f(x) = e^{2x}$ , find  $f'(x)$ .*

# Differentiation

- Oftentimes, finding analytic derivative is hard, but finding numerical derivative is often possible.

## Definition 4

For a function  $f$  and a small constant  $h$ ,

- $f'(x) \approx \frac{f(x+h)-f(x)}{h}$  (forward difference formula)
- $f'(x) \approx \frac{f(x)-f(x-h)}{h}$  (backward difference formula)
- $f'(x) \approx \frac{f(x+h)-f(x-h)}{2h}$  (centered difference formula)



# Integration

## Definition 5 (integration)

**Integration** is the computation of an integral, which is a reverse operation of differentiation up to an additive constant.

## Definition 6 (integral or antiderivative)

Let's say a function  $f$  is a derivative of  $g$ , or  $g'(x) = f(x)$ , then we say  $g$  is an **integral** or, **antiderivative** of  $f$ , written as  $g(x) = \int f(x)dx + C$ , where  $C$  is a integration constant.

### Remark 3

The followings are popular integrals.

- For  $p \neq -1$ ,  $f(x) = x^p \Rightarrow \int f(x)dx = \frac{1}{p+1}x^{p+1} + C$  (polynomial)
- $f(x) = \frac{1}{x} \Rightarrow \int f(x)dx = \log(x) + C$  (fraction)
- $f(x) = e^x \Rightarrow \int f(x)dx = e^x + C$  (exponential)
- $f(x) = \frac{g'(x)}{g(x)} \Rightarrow \int f(x)dx = \log(g(x)) + C$  (See Theorem 4 above)

### Exercise 3

Derive  $\int f'(x)g(x) dx = f(x)g(x) - \int f(x)g(x)' dx$ . (Hint: Use Theorem 2 above.)

## Exercise 4

Find  $\int x e^x dx$ , and evaluate  $\int_0^1 x e^x dx$ . (Hint: Use Exercise 3 above.)



## II. Numerical methods for finding a root

## About - solving an equation

- For the rest of this section, we consider a nonlinear and differentiable (thus, continuous) function  $f : \mathbb{R} \rightarrow \mathbb{R}$ , we aim to find a point  $x^* \in \mathbb{R}$  such that  $f(x^*) = 0$ . We call such  $x^*$  as a *solution* or a *root*.

# 1. Bisection Method

- The *bisection* method aims to find a very short interval  $[a, b]$  in which  $f$  changes a sign.
- Why? Changing a sign from  $a$  to  $b$  means the function crosses the  $\{y = 0\}$ -axis (in other words  $x$ -axis) at least once. It means that  $x^*$  such that  $f(x^*) = 0$  is within this interval. Since  $[a, b]$  is a very short interval, We may simply say  $x^* = \frac{a+b}{2}$ .

## Definition 7 (sign function)

$\text{sgn}(\cdot)$  is called a *sign function* that returns 1 if the input is positive, -1 if negative, and 0 if zero.

- Let  $tol$  be the maximum allowable length of the **short interval** and an initial interval  $[a, b]$  be such that  $sgn(f(a)) \neq sgn(f(b))$ .
- The **bisection algorithm** is the following.

```
1: while  $((b - a) > tol)$  do
2:    $m = \frac{a+b}{2}$ 
3:   if  $sgn(f(a)) = sgn(f(m))$  then
4:      $a = m$ 
5:   else
6:      $b = m$ 
7:   end
8: end
```

- At each **iteration**, the interval length is halved. As soon as the interval length becomes smaller than  $tol$ , then the algorithm stops.



## 2. Newton Method

- The bisection technique makes no use of the function values other than their signs, resulting in slow but sure convergence.
- More rapid convergence can be achieved by using the function values to obtain a more accurate approximation to the solution, at each iteration.
- Newton method is a method that use both the function value and derivative value.
- Newton method approximates the function  $f$  near  $x_k$  by the tangent line at  $f(x_k)$ .

1:  $x_0 =$  initial guess

2: for  $k = 0, 1, 2, \dots$

3:      $x_{k+1} = x_k - f(x_k)/f'(x_k)$

4:     break if  $|x_{k+1} - x_k| < tol$

5: end

### 3. Fixed point theorem

#### Definition 8 (Fixed point)

For a function  $f(\cdot)$ ,  $x^*$  is called a fixed point if  $f(x^*) = x^*$  holds.

#### Remark 4

- For example,  $x^* = 2$  is a fixed point for  $f(x) = x^2 - 3x + 4$ .
- Not all functions have fixed points. For example,  $f(x) = x + 1$ .
- In graphical terms, a fixed point  $x$  means the point  $(x, f(x))$  is on the line  $y = x$ .
- In other words the graph of  $f$  has a point in common with that line.

## Theorem 1 (contraction mapping theorem)

*Let  $x_0$  to be an arbitrary point, and let  $x_{k+1} = f(x_k)$  for  $k \geq 0$ . Under certain condition of  $f$ , the sequence of  $\{x_n\}$  converges to  $x^*$  such that  $f(x^*) = x^*$ .*

- 1:  $x_0 =$  initial guess
- 2: for  $k = 0, 1, 2, \dots$
- 3:      $x_{k+1} = x_k - f(x_k)/f'(x_k)$
- 4:     break if  $|x_{k+1} - x_k| < tol$
- 5: end

- Consider  $f(x) = 1 + 1/x$ .
- Its solution to  $f(x^*) = x^*$  can be solved by  $x = 1 + 1/x$ , or  $x^2 - x - 1 = 0$ .
- In other words,  

$$x^* = \frac{1 \pm \sqrt{5}}{2} \approx 1.618 \text{ or } -0.618$$

```
f <- function(x) {
  return(1+1/x)
}
tol <- 10^(-5)
x_now <- 0.1
```

```
repeat{
  x_next <- f(x_now)
  if (abs(x_next-x_now) < tol) {
    break
  }
  x_now <- x_next
  print(x_next)
}
```

```
## [1] 11
## [1] 1.090909
## [1] 1.916667
## [1] 1.521739
## [1] 1.657143
## [1] 1.603448
## [1] 1.623656
## [1] 1.615894
## [1] 1.618852
## [1] 1.617722
## [1] 1.618153
## [1] 1.617988
## [1] 1.618051
## [1] 1.618027
```

## Exercise 5

*Write a python code that does the exactly same thing as the above code block.*

## Summary

- The above mentioned root-finding numerical methods share a few common properties.
  - 1 It is characterized as a *iterative process* (such as  $x_0 \rightarrow x_1 \rightarrow x_2 \rightarrow \dots$ ).
  - 2 In each *iteration*, the current candidate for the solution *gets closer* to the true value.
  - 3 It converges. That is, it is theoretically reach the *exact value* up to tolerance.
- Many iterative numerical methods share the properties above.
- The famous back propagation in deep neural network is also motivated by Newton method.
- Major algorithms for dynamic programming are called *policy iteration* and *value iteration* that also share the properties above.



### III. Matrix algebra



# Matrix multiplication

## Exercise 6

*Solve the followings.*

$$(.6 \quad .4) \begin{pmatrix} .7 & .3 \\ .5 & .5 \end{pmatrix} =$$

## Exercise 7

What is  $P^2$ ?

$$P = \begin{pmatrix} .7 & .3 \\ .5 & .5 \end{pmatrix}$$

# Solution to system of linear equations

## Exercise 8

*Solve the followings.*

$$(\mathbf{v}_1 \quad \mathbf{v}_2) \begin{pmatrix} .7 & .3 \\ .5 & .5 \end{pmatrix} = (\mathbf{v}_1 \quad \mathbf{v}_2)$$

$$\mathbf{v}_1 + \mathbf{v}_2 = 1$$

## Exercise 9

*Solve the following system of equations.*

$$x = y$$

$$y = 0.5z$$

$$z = 0.6 - 0.4x$$

$$x + y + z = 1$$

## Exercise 10

*Solve the following system of equations.*

$$(\mathbf{v}_0 \quad \mathbf{v}_1 \quad \mathbf{v}_2) \begin{pmatrix} -2 & 2 \\ 3 & -5 & 2 \\ & 3 & -3 \end{pmatrix} = (0 \quad 0 \quad 0)$$

$$\mathbf{v}_0 + \mathbf{v}_1 + \mathbf{v}_2 = 1$$

## Exercise 11

Solve the following system of equations.

$$(\mathbf{v}_1 \quad \mathbf{v}_2 \quad \mathbf{v}_3 \quad \mathbf{v}_4) \begin{pmatrix} .7 & .3 \\ .5 & .5 \\ & .6 & .4 \\ & .3 & .7 \end{pmatrix} = (\mathbf{v}_1 \quad \mathbf{v}_2 \quad \mathbf{v}_3 \quad \mathbf{v}_4)$$

$$\mathbf{v}_1 + \mathbf{v}_2 + \mathbf{v}_3 + \mathbf{v}_4 = 1$$

$$\mathbf{v}_1 + \mathbf{v}_2 = a$$

## Exercise 12

Solve following and express  $\mathbf{v}_i$  for  $i = 0, 1, 2, \dots$

$$\begin{aligned}
 \mathbf{v}_0 + \mathbf{v}_1 + \mathbf{v}_2 + \dots &= 1 \\
 0.02\mathbf{v}_0 + 0.02\mathbf{v}_1 + 0.02\mathbf{v}_2 + \dots &= \mathbf{v}_0 \\
 0.98\mathbf{v}_0 &= \mathbf{v}_1 \\
 0.98\mathbf{v}_1 &= \mathbf{v}_2 \\
 0.98\mathbf{v}_2 &= \mathbf{v}_3 \\
 \dots &= \dots
 \end{aligned}$$





## IV. Series and others

## Exercise 13 (Infinite geometric series)

*Simplify the following. When  $|r| < 1$ ,  $S = a + ar + ar^2 + ar^3 + \dots$*

## Exercise 14 (Finite geometric series)

*Simplify the following. When  $r \neq 1$ ,  $S = a + ar + ar^2 + ar^3 + \dots + ar^{n-1}$*

## Exercise 15 (Power series)

*Simplify the following. When  $|r| < 1$ ,  $S = r + 2r^2 + 3r^3 + 4r^4 + \dots$*

# Formulation of time varying function

## Exercise 16

*During the first hour ( $0 \leq t \leq 1$ ),  $\lambda(t)$  increases linearly from 0 to 60. After the first hour,  $\lambda(t)$  is constant at 60. Draw plot for  $\lambda(t)$  and express the function in math form.*



"Man can learn nothing unless he proceeds from the known to the unknown. - Claude Bernard"