

MATH7501 Practical 12 (Week 13), Semester 1-2021

Topic: Revision

Author: Aminath Shausan

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Pre-Tutorial Activity

- Students must have familiarised themselves with units 1 and 10 contents of the reading materials for MATH7501

Resources

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Q 1: Taylor Series Approximation

a) Find Taylor series approximation for $\ln(1+x^2)$ about $x = 0$

As the expansion is around 0, this approximation is given by the Maclaurine series :

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

First consider the Maclaurine series approximation for $g(x) =$

$\ln(1+x)$. The first few derivatives of $g(x)$ at $x = 0$ are given below.

$$g'(0) = \frac{1}{1+x} \Big|_{x=0} = 1$$

$$g''(0) = \frac{-1}{(1+x)^2} \Big|_{x=0} = -1$$

$$g^{(3)}(0) = \frac{2}{(1+x)^3} \Big|_{x=0} = 2$$

$$g^{(4)}(0) = \frac{-6}{(1+x)^4} \Big|_{x=0} = -6$$

$$g^{(5)}(0) = \frac{42}{(1+x)^5} \Big|_{x=0} = 24$$

$$g^{(6)}(0) = \frac{-120}{(1+x)^6} \Big|_{x=0} = -120$$

....

$$g^{(n)}(0) = \frac{(-1)^{n+1} (n-1)!}{(1+x)^n} \Big|_{x=0} = (-1)^{n+1} (n-1)!$$

Thus

$$\begin{aligned} g(x) = \ln(1+x) &\approx \sum_{n=0}^{\infty} \frac{(-1)^{n+1} (n-1)!}{n!} x^n \\ &= \sum_{n=0}^{\infty} \frac{(-1)^{n+1}}{n} x^n \end{aligned}$$

Replacing x with x^2 in the above we get

$$\begin{aligned} f(x) = \ln(1+x^2) &\approx \sum_{n=0}^{\infty} \frac{(-1)^{n+1}}{n} (x^2)^n \\ &= x^2 - \frac{x^4}{2} + \frac{x^6}{3} - \frac{x^8}{4}, \end{aligned}$$

up to order 8. You can use the Series[] function to check this.

(* this computes Taylor series expansion of $f(x)$ at $x=0$ upto order 8*)
Series[Log[1+x^2], {x, 0, 8}]

$$\text{Out[4]= } x^2 - \frac{x^4}{2} + \frac{x^6}{3} - \frac{x^8}{4} + O[x]^9$$

b) Find the values of x for which the series converges.

Let $a_n = \frac{(-1)^{n+1}}{n} x^{2n}$, then by the Ratio test,

the series $\sum_{n=0}^{\infty} \frac{(-1)^{n+1}}{n} (x^2)^n$ converges if $\lim_{n \rightarrow \infty} \left| \frac{a_{n+1}}{a_n} \right| < 1$. Now

$$\begin{aligned} \lim_{n \rightarrow \infty} \left| \frac{a_{n+1}}{a_n} \right| &= \lim_{n \rightarrow \infty} \left| \frac{(-1)^{(n+1)+1}}{n+1} x^{2(n+1)} \times \frac{(-1)^{n+1}}{n} x^{2n} \right| \\ &= \lim_{n \rightarrow \infty} \left| x^2 \frac{n}{n+1} \right| \\ &= |x^2| \end{aligned}$$

Thus for the above series to converge, it requires that

$$|x^2| < 1$$

$$\sqrt{x} < \sqrt{1}$$

$$|x| < 1.$$

Therefore, the series converges for $|x| < 1$.

c) Find Taylor series approximation for $\ln(1+x^2)$ about $x = a$

(* this computes Taylor series expansion of $f(x)$ at $x=a$ upto order 4*)

In[6]:= Series[Log[1+x^2], {x, a, 4}]

$$\begin{aligned} \text{Out[6]= } &\text{Log}[1+a^2] + \frac{2a(x-a)}{1+a^2} + \frac{(1-a^2)(x-a)^2}{(1+a^2)^2} + \\ &\frac{2a(-3+a^2)(x-a)^3}{3(1+a^2)^3} + \frac{(-1+6a^2-a^4)(x-a)^4}{2(1+a^2)^4} + O[x-a]^5 \end{aligned}$$

(*Eg. if the expansion is about $x=1$, then we have*)

In[7]:= Series[Log[1 + x²], {x, 1, 4}]

Out[7]= Log[2] + (x - 1) - $\frac{1}{6}$ (x - 1)³ + $\frac{1}{8}$ (x - 1)⁴ + O[x - 1]⁵

Q 2: Gradient Computation and Taylor Series Approximation

Consider $f(\mathbf{x}) = x_1 + e^{(x_2 - x_1)}$.

a) Find the gradient of $f(\mathbf{x}) : \nabla f(\mathbf{x})$

$$\begin{aligned}\nabla f(\mathbf{x}) &= \left(\frac{\partial}{\partial x_1} f(\mathbf{x}), \frac{\partial}{\partial x_2} f(\mathbf{x}) \right)^T \\ &= (1 - e^{(x_2 - x_1)}, e^{(x_2 - x_1)})^T\end{aligned}$$

b) Use $\nabla f(\mathbf{x})$ to find a Taylor series approximation of $f(\mathbf{x})$ around $\mathbf{x} = (1, 2)$ up to order 1

$$\begin{aligned}f(\mathbf{x}) &\approx f(1, 2) + \nabla f(1, 2) (\mathbf{x} - (1, 2))^T \\ &= 3.7183 + (-1.7183, 2.7183) (x_1 - 1, x_2 - 2)^T \\ &= 3.7183 - 1.7183 (x_1 - 1) + 2.7183 (x_2 - 2)\end{aligned}$$

Q 3. Determinant of a 3 × 3 Matrix

Find the determinant of $A = \begin{pmatrix} 7 & 2 & 1 \\ 0 & 3 & -1 \\ -3 & 4 & -2 \end{pmatrix}$

Using cofactors and expanding along the first row, gives :

$$\begin{aligned}\det(A) &= |A| \\ &= (-1)^{1+1} 7 \begin{vmatrix} 3 & -1 \\ 4 & -2 \end{vmatrix} + (-1)^{1+2} 2 \begin{vmatrix} 0 & -1 \\ -3 & -2 \end{vmatrix} + (-1)^{1+3} 1 \begin{vmatrix} 0 & 3 \\ -3 & 4 \end{vmatrix} \\ &= 7(-6 - (-4)) + 2(0 - (-3)) + 1(0 - (-9)) \\ &= 7(-2) + 2(-3) + 1(9) \\ &= -14 + 6 + 9 \\ &= 1\end{aligned}$$

Q 4. Integration

Find $\int e^x \sin(x) dx$

Need to use integration by parts : $\int u dv =$

$uv - \int v du$. Assuming $u = e^x$ and $dv = \sin(x)$ we get :

$$\int e^x \sin(x) dx = -e^x \cos(x) + \int e^x \cos(x) dx$$

Applying bi-parts for the integral on the right-hand side with $u = e^x$ and $dv = \cos(x)$, we get

$$\int e^x \sin(x) dx = -e^x \cos(x) + \{e^x \sin(x) - \int e^x \sin(x) dx\}$$

$$2 \int e^x \sin(x) dx = -e^x \cos(x) + e^x \sin(x)$$

$$\int e^x \sin(x) dx = \frac{1}{2} e^x [\sin(x) - \cos(x)]$$

Q 5. The Normal Distribution and Moments of Truncated Normal Distribution

Some notations for standard normal random variable

Note that if a random variable X has the standard normal distribution then its mean is 0 and variance is 1. This is written as X as: $X \sim \text{Normal}(0, 1)$.

- the **pdf** of $X \sim \text{Normal}(0, 1)$ is denoted as: $\phi(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}$
- the **cdf** of $X \sim \text{Normal}(0, 1)$ is denoted as: $\Phi(x) = \int_{-\infty}^x \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt$

Cdf and pdf of $X \sim \text{Normal}(\mu, \sigma^2)$ in terms of standard normal random variable

- the **pdf** of $X \sim \text{Normal}(\mu, \sigma^2)$ is given as: $f_X = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$. This can be rearranged to give $f_X(x) = \frac{1}{\sigma} \left(\frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2} \left(\frac{x-\mu}{\sigma} \right)^2} \right) = \frac{1}{\sigma} \phi\left(\frac{x-\mu}{\sigma}\right)$, where $\phi(\cdot)$ is the pdf of $X \sim \text{Normal}(0, 1)$
- the **pdf** of $X \sim \text{Normal}(\mu, \sigma^2)$ is given as: $F_X(x) = \int_{-\infty}^x \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(t-\mu)^2}{2\sigma^2}} dt$. This can be rearranged to give $F_X(x) = \frac{1}{\sigma} \int_{-\infty}^x \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2} \left(\frac{t-\mu}{\sigma} \right)^2} dt = \frac{1}{\sigma} \Phi\left(\frac{x-\mu}{\sigma}\right)$, $\Phi(\cdot)$ is the cdf of $X \sim \text{Normal}(0, 1)$

Truncated Normal Distribution

Let X be a normal random variable with mean μ and variance σ^2

and lies within the interval (a, b) such that $-\infty \leq a < b \leq \infty$. Then X , conditional on $a < X < b$, has a **truncated normal distribution**. Its pdf is given by

$$f_X(x) = \begin{cases} \frac{1}{\sigma} \frac{\phi\left(\frac{x-\mu}{\sigma}\right)}{\Phi\left(\frac{b-\mu}{\sigma}\right) - \Phi\left(\frac{a-\mu}{\sigma}\right)}, & a \leq x \leq b, \sigma > 0 \\ 0, & \text{Otherwise} \end{cases}$$

Here, $\phi(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}$, is the pdf of standard normal random distribution and

$$\Phi(x) = \int_{-\infty}^x \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt,$$

is the cdf of standard normal distribution ($X \sim \text{Normal}(0, 1)$)

Consider now the k -th moment of the truncated normal distribution

$m_k := E[X^k]$, for $a \leq X \leq b$ and $a, b \in [-\infty, \infty]$, (this implies X is truncated normal)
 where $E[X^k]$ denotes the expected value of X^k (the k -th moment of X).

(a) Show that m_k can be given by the recursive formula :

$$m_k = (k-1) \sigma^2 m_{k-2} + \mu m_{k-1} - \sigma \frac{b^{k-1} \phi\left(\frac{b-\mu}{\sigma}\right) - a^{k-1} \phi\left(\frac{a-\mu}{\sigma}\right)}{\Phi\left(\frac{b-\mu}{\sigma}\right) - \Phi\left(\frac{a-\mu}{\sigma}\right)}, \text{ for } k = 1, 2, \dots$$

with $m_{-1} = 0$ and $m_0 = 1$,

where $\phi(\cdot)$ is the standard normal pdf and $\Phi(\cdot)$ is the standard normal cdf.

ANS : Let $n_0 := \Phi\left(\frac{b-\mu}{\sigma}\right) - \Phi\left(\frac{a-\mu}{\sigma}\right)$. Then

$$\begin{aligned} m_k n_0 &= \int_a^b x^k \frac{1}{\sigma} \phi\left(\frac{x-\mu}{\sigma}\right) dx \\ &= \int_a^b x^k \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx \\ &= \int_a^b x^{k-1} \left(\frac{x\sigma}{\sigma}\right) \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx \\ &= \int_a^b \frac{\sigma}{\sqrt{2\pi}} x^{k-1} \left(\frac{x}{\sigma^2}\right) e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx \\ &= \int_a^b \frac{\sigma}{\sqrt{2\pi}} x^{k-1} \left(\frac{x-\mu}{\sigma^2} + \frac{\mu}{\sigma^2}\right) e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx \\ &= \sigma \int_a^b \frac{(x-\mu)}{\sigma^2 \sqrt{2\pi}} x^{k-1} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx + \int_a^b \frac{\mu}{\sigma \sqrt{2\pi}} x^{k-1} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx \\ &= \sigma \left[\frac{-x^{k-1}}{\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \right]_{x=a}^{x=b} + \int_a^b (k-1) \frac{x^{k-2}}{\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx + \int_a^b \frac{\mu}{\sigma \sqrt{2\pi}} x^{k-1} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx, \end{aligned}$$

using bi-parts

$$\begin{aligned} &= \sigma \left[\frac{-b^{k-1}}{\sqrt{2\pi}} e^{-\frac{(b-\mu)^2}{2\sigma^2}} + \frac{a^{k-1}}{\sqrt{2\pi}} e^{-\frac{(a-\mu)^2}{2\sigma^2}} \right] + \\ &\quad (k-1) \sigma^2 \int_a^b \frac{x^{k-2}}{\sigma \sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx + \mu \int_a^b \frac{1}{\sigma \sqrt{2\pi}} x^{k-1} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx \\ &= \sigma \left[-b^{k-1} \phi\left(\frac{b-\mu}{\sigma}\right) + a^{k-1} \phi\left(\frac{a-\mu}{\sigma}\right) \right] + (k-1) \sigma^2 \int_a^b \frac{x^{k-2}}{\sigma} \phi\left(\frac{x-\mu}{\sigma}\right) dx + \\ &\quad \mu \int_a^b \frac{1}{\sigma} x^{k-1} \phi\left(\frac{x-\mu}{\sigma}\right) dx, \text{ using the pdf of standard normal distribution.} \end{aligned}$$

Thus

$$\begin{aligned} m_k &= \frac{\sigma \left[-b^{k-1} \phi\left(\frac{b-\mu}{\sigma}\right) + a^{k-1} \phi\left(\frac{a-\mu}{\sigma}\right) \right]}{n_0} + \\ &\quad (k-1) \sigma^2 \int_a^b \frac{x^{k-2}}{n_0 \sigma} \phi\left(\frac{x-\mu}{\sigma}\right) dx + \mu \int_a^b \frac{1}{n_0 \sigma} x^{k-1} \phi\left(\frac{x-\mu}{\sigma}\right) dx. \end{aligned}$$

Observing that the integrals in the second and third terms are lower order moments and reorganising, we obtain the result.

b) Find a recursive formula for the k -th moment of $X \sim \text{Normal}(\mu, \sigma^2)$ where X lies in the interval $(-\infty, \infty)$.

$$\begin{aligned}
 m_k = E[X^k] &= \int_{-\infty}^{\infty} x^k \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx, \\
 &= \int_{-\infty}^{\infty} x^{k-1} (x - \mu + \mu) \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx \\
 &= \int_{-\infty}^{\infty} x^{k-1} (x - \mu) \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx + \int_{-\infty}^{\infty} \mu x^{k-1} \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx \\
 &= \left[\frac{-\sigma x^{k-1}}{\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \right]_{x=-\infty}^{x=\infty} + \int_{-\infty}^{\infty} (k-1) x^{k-2} \frac{\sigma}{\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx + \\
 &\quad \int_{-\infty}^{\infty} \mu x^{k-1} \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx, \text{ by applying integration by parts} \\
 &= 0 + (k-1) \sigma^2 \int_{-\infty}^{\infty} x^{k-2} \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx + \mu m_{k-1} \\
 &= (k-1) \sigma^2 m_{k-2} + \mu m_{k-1}, \quad k > 2.
 \end{aligned}$$