

# Homework1

## Problem 1

For a uniform distribution, the expectation value is:

$$\langle X \rangle = \int_a^b X \frac{1}{b-a} dX = \frac{a+b}{2} = 1$$

The variance in this case is:

$$\begin{aligned} V(X) &= \langle X^2 \rangle - \langle X \rangle^2 \\ &= \int_a^b X^2 \frac{1}{b-a} dX - \langle X \rangle^2 \\ &= \frac{1}{3} (a^2 + ab + b^2) - \langle X \rangle^2 = 0.16 \end{aligned}$$

So, we can calculate:

$$\begin{cases} a+b=2 \\ b>a \\ a^2+ab+b^2=3.48 \end{cases} \rightarrow \begin{cases} a=0.30718 \\ b=1.69282 \end{cases}$$

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## Problem 2

$$\begin{aligned} \phi(\lambda) &= \langle e^{\lambda x} \rangle = \left\langle \sum_{n=0}^{\infty} \frac{(\lambda X)^n}{n!} \right\rangle \\ &= \sum_{n=0}^{\infty} \frac{\lambda^n}{n!} \langle X^n \rangle \end{aligned}$$

For a variable with Gaussian distribution, the MGF could be calculated by:

$$\phi(\lambda|\mu, \sigma) = \frac{1}{\sqrt{2\pi\sigma^2}} \int_{-\infty}^{\infty} e^{\lambda x} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx$$

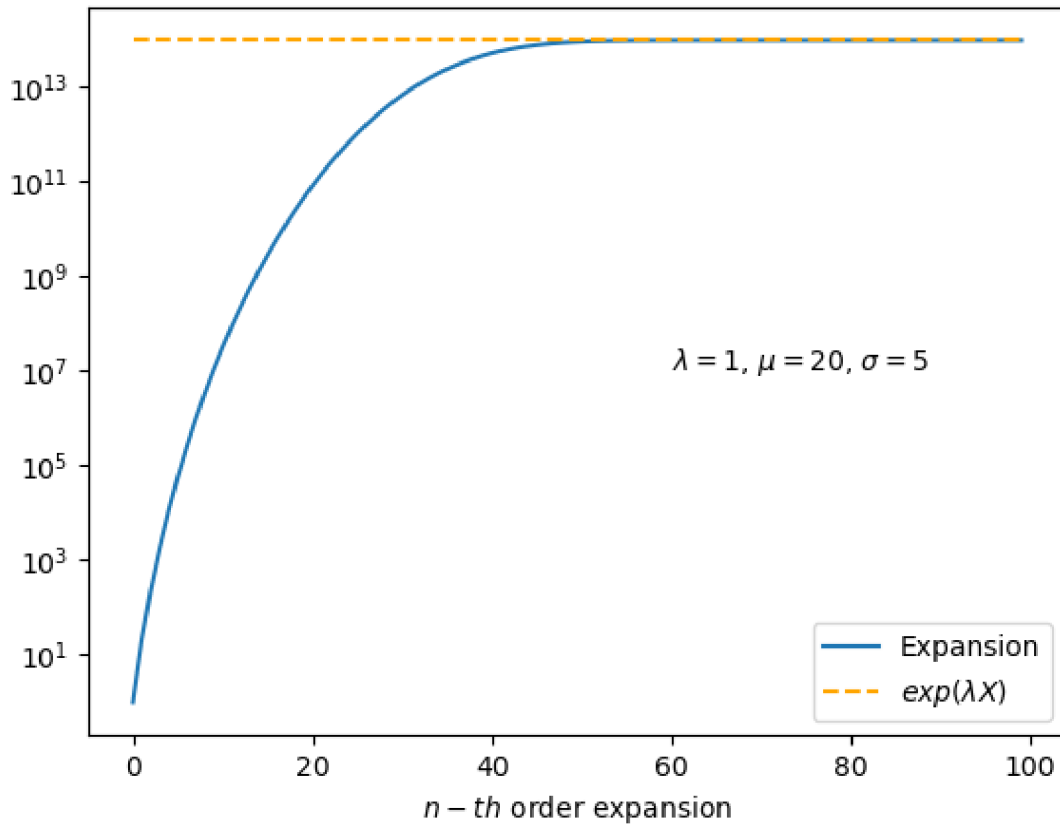
For the index term:

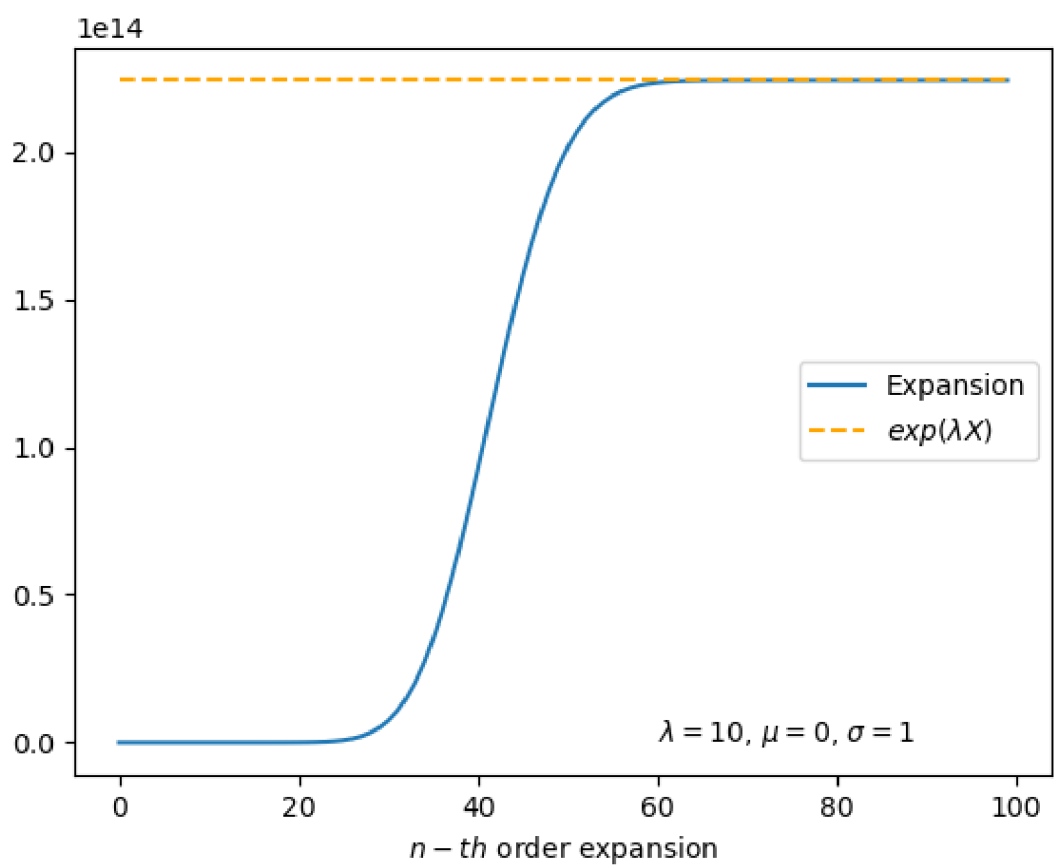
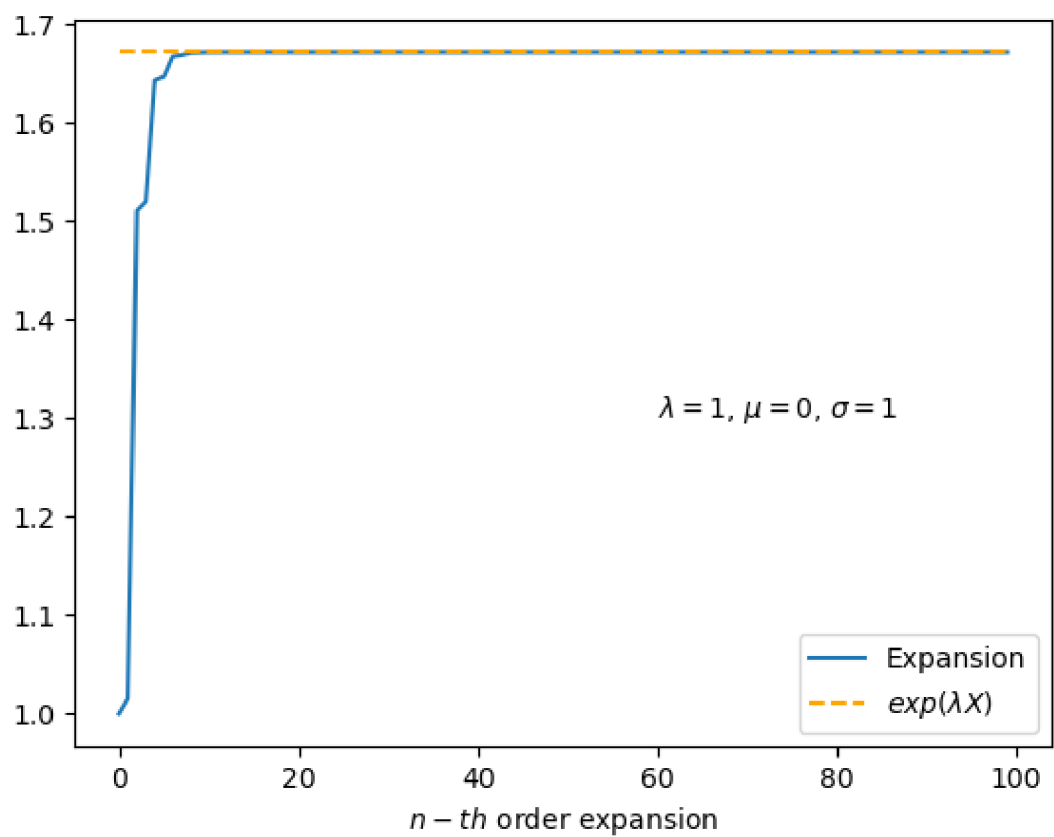
$$\begin{aligned}
-\frac{(x-\mu)^2}{2\sigma^2} + \lambda x &= -\frac{1}{2\sigma^2} (x^2 - 2\mu x + \mu^2) + \lambda x \\
&= -\frac{1}{2\sigma^2} x^2 + \frac{\mu}{\sigma^2} x - \frac{\mu^2}{2\sigma^2} + \lambda x \\
&= -\frac{1}{2\sigma^2} x^2 + \left( \frac{\mu}{\sigma^2} + \lambda \right) x - \frac{\mu^2}{2\sigma^2} \\
&= -\frac{1}{2\sigma^2} \left[ x^2 - (2\mu + 2\lambda\sigma^2)x + (\mu + \lambda\sigma^2)^2 \right] + \frac{1}{2\sigma^2} (\mu + \lambda\sigma^2)^2 - \frac{\mu^2}{2\sigma^2}
\end{aligned}$$

So, the final integral becomes:

$$\begin{aligned}
\phi(\lambda|\mu, \sigma) &= \frac{1}{\sqrt{2\pi\sigma^2}} \int_{-\infty}^{\infty} e^{\lambda x} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx \\
&= e^{\frac{1}{2\sigma^2} (\mu + \lambda\sigma^2)^2 - \frac{\mu^2}{2\sigma^2}} = \exp\left(\mu\lambda + \frac{1}{2}\sigma^2\lambda^2\right)
\end{aligned}$$

### Problem 3





## Problem 4

Suppose  $\langle \hat{X} \rangle = \mu$ ,  $\langle (\hat{X} - \mu)^2 \rangle = \sigma^2$ , and  $f(\hat{X})$  is any function. Suppose  $\hat{X} \approx \mu$ , then we have:

$$f(\hat{X}) = f(\mu) + f'(\mu)(\hat{X} - \mu) + \frac{f''(\mu)}{2!}(\hat{X} - \mu)^2 + \dots + \frac{f^{(n)}(\mu)}{n!}(\hat{X} - \mu)^n$$
$$\langle f(\hat{X}) \rangle = f(\mu) + f'(\mu)\langle (\hat{X} - \mu) \rangle + \frac{f''(\mu)}{2!}\langle (\hat{X} - \mu)^2 \rangle + \dots + \frac{f^{(n)}(\mu)}{n!}\langle (\hat{X} - \mu)^n \rangle$$

For a non-linear function, all the derivatives can not be zero simultaneously. For a specific example, let's say  $\hat{Y} = \hat{X}^2$ , then,

$$\langle \hat{Y} \rangle = \mu^2 + 0 + \sigma^2 = \mu^2 + \sigma^2$$