

# The Normal Distribution Unit

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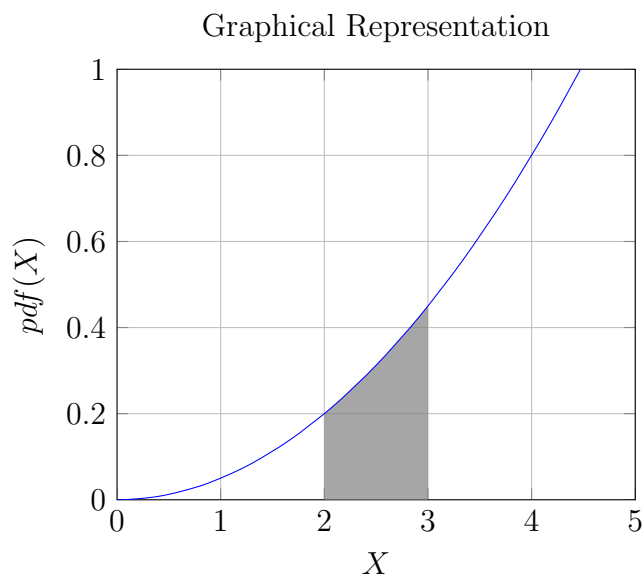
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## 1 Introduction

### PDF (Probability Density Function)

The probability per unit change of the variable at a particular point.

Instead of the dependent variable being probability  $P(X)$  it's replaced with  $pdf(X)$ .



The shaded area represents  $P(2 < X < 3)$ . NOTE:  $\leq$  and  $<$  are the same

### Terms

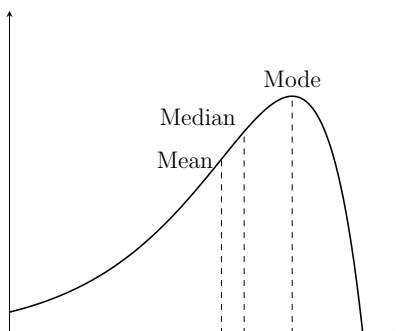
Unitary Condition: Total Probability = 1

Continuous Random Variable: No breaks in continuity (All values exist within its range)

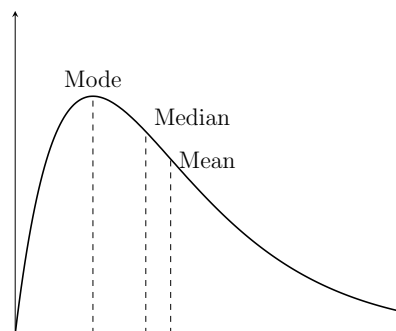
Unimodal: One mode/median/mean on the curve

Bimodal: Two modes/medians/means on the curve

Negatively Skewed / Skewed Left



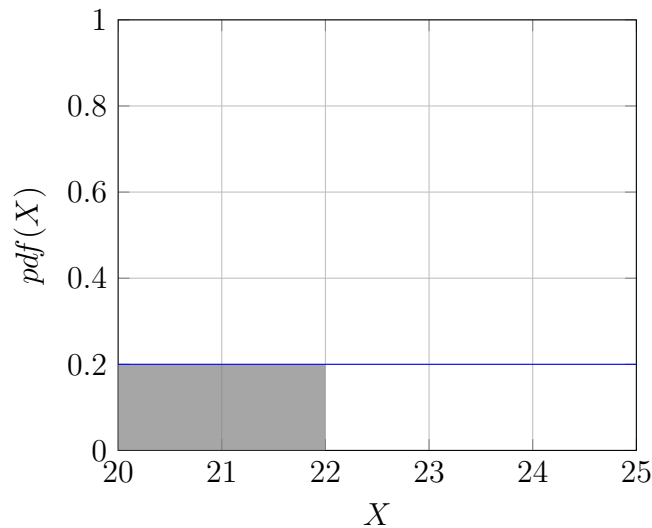
Positively Skewed / Skew Right



## Uniform Distribution

Probability Density remains constant.

Graphical Representation



In this example, we are given  $a = \frac{1}{5}$ .

$$P(20 < X < 22) = 2 \cdot a = 2 \cdot \frac{1}{5} = \frac{2}{5}$$

## Probability Density

$$P(X < a) = ke^{-ka}$$

$\mu$  : average time

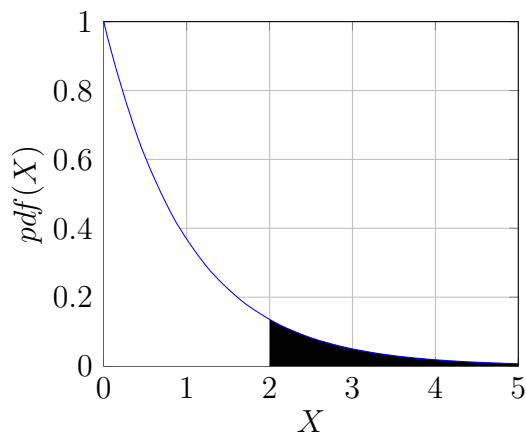
$k = \frac{1}{\mu}$  : # of events per unit of time, ex. #per hr/min/sec

## Exponential Distribution

Used to describe or find the probability related to time taken between 2 consecutive events that occurred.

$$P(X < x) = 1 - e^{-kx} \text{ WHITE}$$

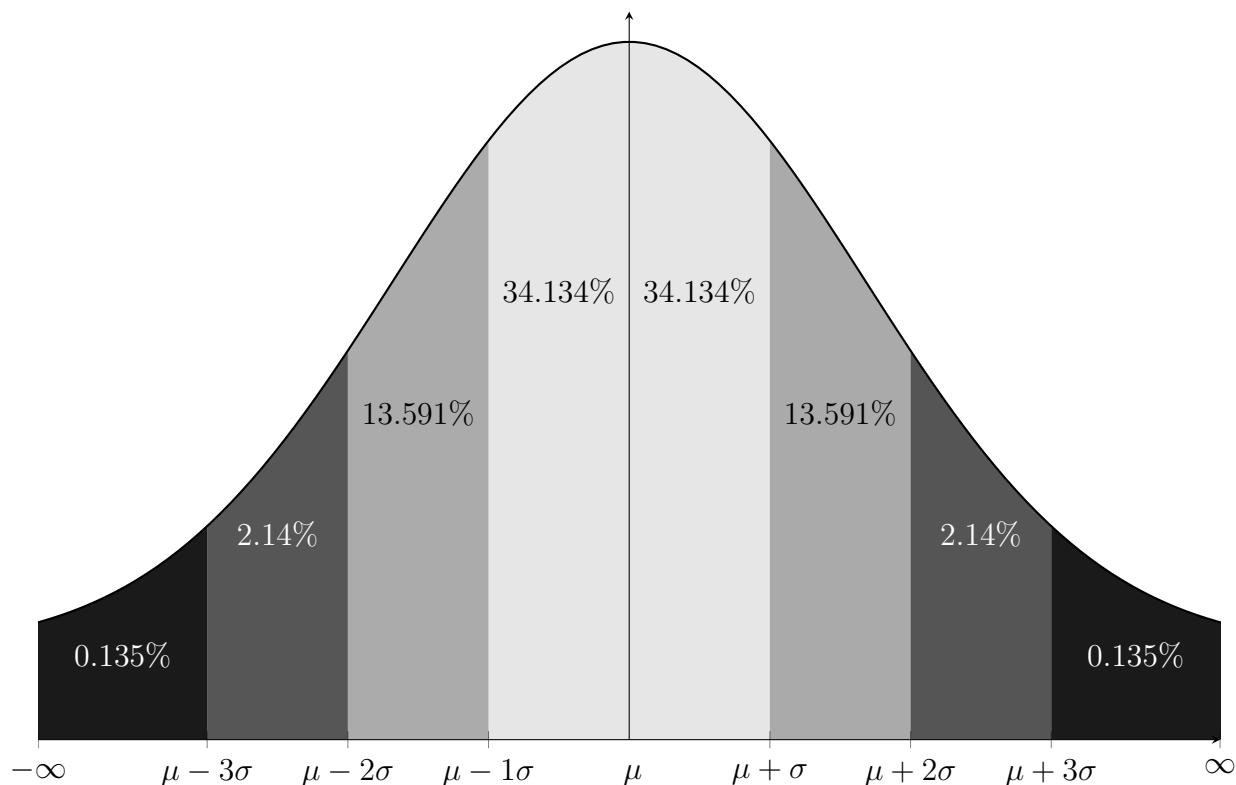
$$P(X > x) = e^{-kx} \text{ BLACK}$$



Example: if  $x = 8, \mu = 10mins$

$$P(X < 8) = 1 - e^{-\frac{1}{10} \cdot 8} \approx 0.55mins$$

## 2 Normal Distribution



### Properties of Normal Distribution

General Properties: Bell Shape, Symmetric, origin represents  $\mu$ , total area under curve = 1

Distribution Type: Continuous Probability Distribution

y-axis represents pdf & area under curve represents probability

Formula:

$$pdf(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$$

$x$  : random variable

$\mu$  : mean

$\sigma$  : Standard Deviation

Exponent  $\left(\frac{x-\mu}{\sigma}\right)^2$  Properties:

When  $x = \mu$ ,  $f(x)$  has the maximum value. The greater  $|x - \mu|$  is, the less  $f(x)$  is.

### Probability for Standard Normal Distribution

Each area interval of  $\mu \pm Z\sigma$  ( $Z = \{0, \pm 1, \pm 2, \pm 3\}$ ) represents a constant percentage which is the probability.

$$P(a \leq x \leq b) = \text{Sum of area intervals a to b} = \int_a^b \frac{e^{\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}}{\sigma\sqrt{2\pi}} dx \text{ only to check answer}$$

**Using Z-score**  $Z = \frac{x-\mu}{\sigma}$

If Z-score isn't whole:  $x = \mu + Z(\sigma)$

Using Z-score table to find Z. Using positive table for positive Z-score and vice versa. Row represents hundredth digits, add the row and the column to find desired Z-score.

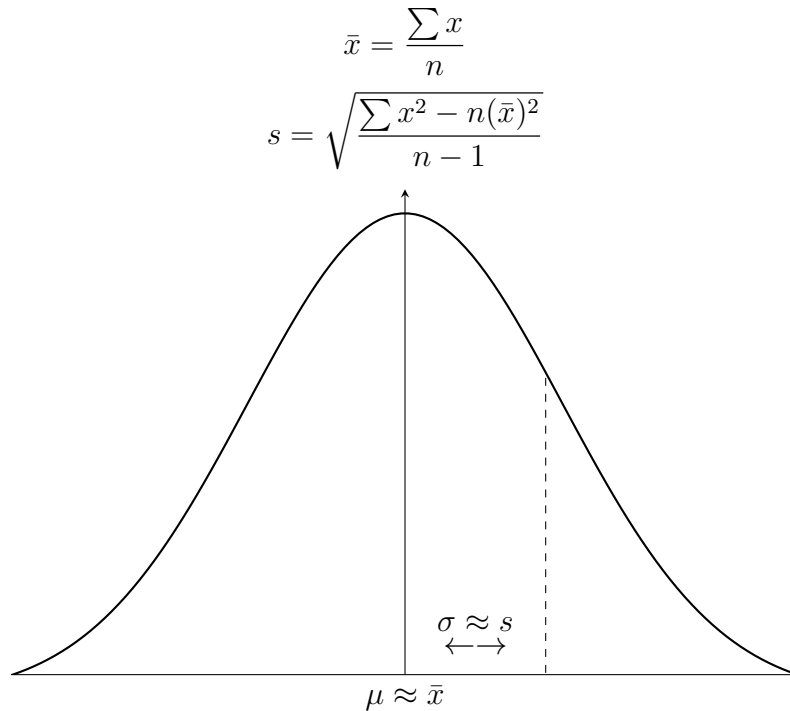
$$P(X < x) = P\left(Z < \frac{x - \mu}{\sigma}\right)$$

$$P(Z < -x) = 1 - P(Z < x)$$

### 3 Normal Sampling and Modelling

#### Normal Sampling

1. The distribution of frequencies in the sample data tends to follow the bell-shaped curve as the underlying distribution.
2. The sample mean ( $\bar{x}$ ) estimates the population mean ( $\mu$ )
3. The sample standard deviation ( $s$ ) estimates the population standard deviation ( $\sigma$ )
4. The larger the sample of a normal population, the more reliable the data will be in reflecting the underlying population



#### Normal Modelling

Being sensible:  $\mu \pm Z\sigma$  shouldn't create any unreasonable data such as negative time, beyond capacity, etc.

Discrete data can sometimes be modelled by a normal distribution. The standard normal curve can be used to approximate the area under the curve which can be used to solve problems involving probability.

#### Continuity Correction

A continuity correction factor is used when you use a continuous probability distribution to approximate a discrete probability distribution.

- $P(X \leq a) \approx P(X < a + 0.5)$
- $P(X < a) \approx P(X < a - 0.5)$
- $P(X \geq a) \approx P(X > a - 0.5)$
- $P(X > a) \approx P(X > a + 0.5)$
- $P(X = a) \approx P(a - 0.5 < X < a + 0.5)$

Example:

$$P(X = 25) \rightarrow P(24.5 < X < 25.5)$$

## 4 Normal Approximation

REMEMBER the binomial distribution is perfectly symmetric if  $p = \frac{1}{2}$ , and has some skewness when not. The normal approximation works best when  $p$  is close to  $\frac{1}{2}$ , and when  $n$  is large.

$$P(X \leq x) = \sum_{k=0}^x \binom{n}{k} p^k q^{n-k}$$

The normal approximation is reasonable if both  $np \geq 5$  and  $n(1-p) \geq 5$ .

For a binomial random variable  $X$ :

$$\begin{aligned}\mu &= np \\ \sigma^2 &= np(1-p) = npq\end{aligned}$$

### Continuity Correction

Desired Probability	Normal Approximation
$P(X \geq x)$	$P(Z > \frac{(x-0.5)-\mu}{\sigma})$
$P(X > x)$	$P(Z > \frac{(x+0.5)-\mu}{\sigma})$
$P(X \leq x)$	$P(Z < \frac{(x+0.5)-\mu}{\sigma})$
$P(X < x)$	$P(Z < \frac{(x-0.5)-\mu}{\sigma})$

## 5 Repeated Sampling and Hypothesis Testing

### Repeated Sampling

When repeated samples of the same size are drawn from a normal population, the sample means will be normally distributed with a mean equal to the population mean ( $\mu_{\bar{x}} = \mu$ ). The distribution of sample means will be where  $n$  is the sample size:

$$\sigma_{\bar{x}} = \frac{\sigma}{\sqrt{n}}$$

To find a  $Z$ -score for a sample mean:

$$P(\bar{x} < x) = P(Z < \frac{(x - 0.5) - \mu_{\bar{x}}}{\sigma_{\bar{x}}})$$

### Hypothesis Testing

A hypothesis test consists of a null hypothesis  $H_0$  and an alternative hypothesis  $H_1$ . The null hypothesis is being challenged is for example  $H_0: \mu = 45$ . While we suspect  $H_1: \mu < 45$ . To obtain a sample mean  $\mu_{\bar{x}} = 44.5$  is very rare, and getting such will prove  $\mu \neq 45$ . Our next step is to establish a decision rule. The significance level  $\alpha$  is the probability threshold for us to determine if observed results are rare enough to justify rejecting  $H_0$ . For example, if  $\alpha = 0.05$ , we are willing to be wrong 5% of the time. We are given  $\sigma = 2, n = 30$ . To begin the test, we assume  $H_0$  is true:

$$\begin{aligned}P(\bar{x} < \mu_{\bar{x}}) &= P(Z < \frac{\mu_{\bar{x}} - \mu}{\sigma_{\bar{x}}}) \\ P(\bar{x} < 44.5) &= P(Z < \frac{44.5 - 45}{0.365}) = P(Z < -1.37) = 0.0835\end{aligned}$$

To conclude, we compare our answer with  $\alpha$ . If the probability is greater than the significance level we accept  $H_0$ , else we accept  $H_1$ . In this case  $0.0835 > \alpha$ , therefore, the evidence isn't sufficient to refute our claim.