

# Ch.5: Continuous Random Variables

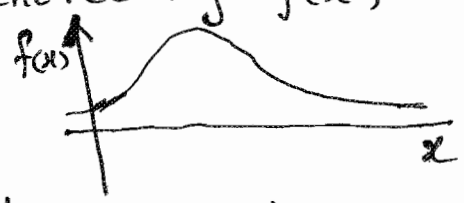
We have already discussed the discrete random variables  $\dashrightarrow$  Binomial Distribution



Continuous R.V.  $\dashrightarrow$  Normal Distribution

- Continuous R.V. can assume any numerical value on an interval
  - The time needed to finish a certain task
  - Temperatures
  - Weights

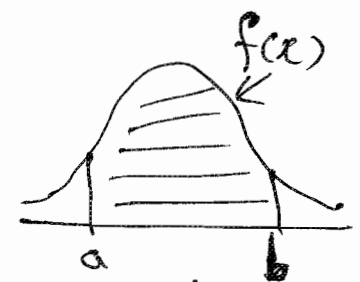
• The graphical form of the probabilities of a continuous r.v.  $X$  is a smooth curve, denoted by  $f(x)$



$f(x)$  is called either

- Probability density function (PDF)
- or frequency function
- Probability distribution

•  $P(a < X < b) =$  Area under  $f(x)$  between  $a$  and  $b$



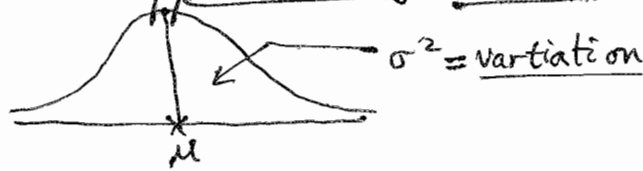
•  $P(X = a) = 0$  [No probability at any single point]

•  $P(a \leq X \leq b) = P(a < X < b)$   
[No Prob.'s at end points]

# Normal Distribution

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- It is well-known distribution, called bell-shaped distribution.
- Many data follow an approximately normal distribution



- The probability density function (pdf) of  $X \sim N(\mu, \sigma^2)$

is

$$f(x) = \frac{1}{\sqrt{2\pi} \sigma} e^{-\frac{1}{2\sigma^2}(x-\mu)^2}, \quad -\infty < x < \infty$$

where  $\mu$  = mean of the normal r.v.  $X$

$\sigma$  = standard deviation

$\pi$  = 3.1416...

$e$  = 2.71828...

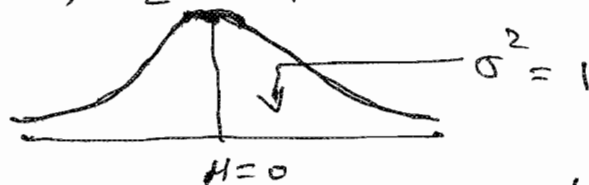
- The pdf of  $X \sim N(\mu, \sigma^2)$  depends on 2 parameters

- $\mu$  = mean of  $X$
- $\sigma^2$  = variance of  $X$

- If  $\mu=0$  and  $\sigma^2=1$ , we will have standard normal distribution

$$f(z) = \frac{1}{\sqrt{e\pi}} e^{-\frac{1}{2}z^2}, \quad -\infty < z < \infty$$

In this case,  $Z \sim N(0, 1)$

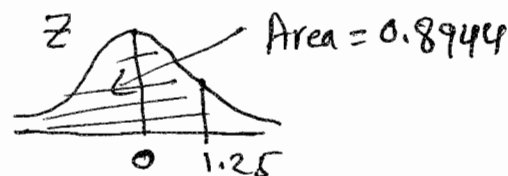


Standard Normal Distribution

# Normal Table

Z	0	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09
0.0										
0.1										
0.2										
0.3										
⋮										
⋮										
1.0	0.8413									
1.1										
1.2	0.8849	0.8869	0.8888	0.8907	0.8925	0.8944	0.8962	0.8980	0.8997	0.9015
⋮										
⋮										
⋮										
3.4										

•  $P(Z \leq 1.25) = T(1.25) = 0.8944$



• The Normal distribution is symmetric

$$P(Z > 1.25) = P(Z < -1.25)$$

$$P(-1 < Z < 0) = P(0 < Z < 1)$$



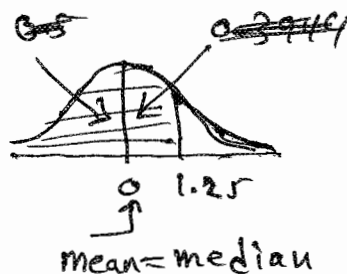
• 
$$P(-1.25 < Z < 1.25) = T(1.25) - T(-1.25)$$
$$= 0.8944 - 0.1056$$
$$= \boxed{0.7888}$$

• 
$$P(1 < Z < 1.25) = T(1.25) - T(1)$$
$$= 0.8944 - 0.8413$$
$$= \boxed{0.0531}$$

$$\begin{aligned}
 \bullet P(Z > 1.25) &= 1 - T(1.25) \\
 &= 1 - 0.8944 \\
 &= \boxed{0.1056}
 \end{aligned}$$

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~~$$\begin{aligned}
 P(Z < 1.25) &= 0.8944 \\
 &= \boxed{0.8944}
 \end{aligned}$$~~



Example 1: The grades of ~~STAT.101~~ STAT.101 are ~~normal~~ normally distributed with mean 70 and standard deviation 10.

What is the probability that a student can get ~~more than~~ more than 90?

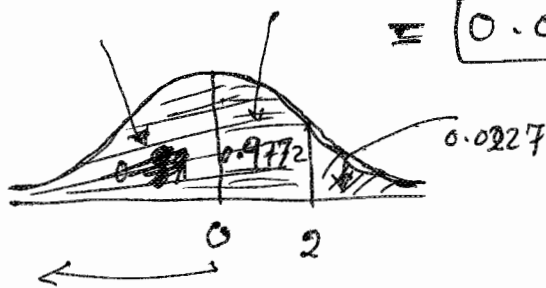
Solution: Need  $P(X > 90)$

(1) Write  $\mu = 70, \sigma = 10$

(2) Change  $X \longrightarrow Z$

$$x = 90 \longrightarrow z = \frac{90 - 70}{10} = [2]$$

$$\begin{aligned}
 (3) \text{ Find } P(Z > 2) &= 1 - T(2) \\
 &= 1 - [1 - 0.9772] \\
 &= \cancel{1 - 0.9772} \\
 &= \boxed{0.0228}
 \end{aligned}$$



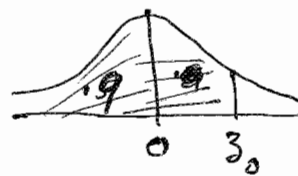
$$\Rightarrow P(X > 90) = P(Z > 2) = \boxed{0.0228}$$

## Percentiles of $N(0,1)$

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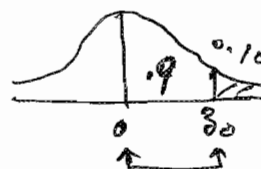
- If  $P(Z \leq z_0) = 0.90$ , find  $z_0$ . Note  $z_0 = 90^{\text{th}}$  Percentile

From Normal table  $P(\text{~~z~~ } Z \leq 1.28)$   
 $= 0.8997$   
 $\approx 0.90$



$$\Rightarrow z_0 = 1.28$$

- If  $P(Z > z_0) = 0.10$ , find  $z_0$



$$\Rightarrow z_0 = 1.28$$

Back to Example 1: Find the grade  $x_0$  ~~that~~ such that 90% of grades fall below  $x_0$  and 10% fall above  $x_0$ . That is,

$$P(X \leq x_0) = 0.90$$

$$P\left(Z \leq \frac{x_0 - 70}{10}\right) = 0.90$$

$$\Rightarrow \frac{x_0 - 70}{10} = 90^{\text{th}} \text{ percentile of } N(0,1) = 1.28$$

$$\frac{x_0 - 70}{10} = 1.28$$

$$\begin{aligned} \Rightarrow x_0 &= 70 + (1.28)(10) \\ &= 70 + 12.8 \\ &= \boxed{82.8} \end{aligned}$$

# Approximating a Binomial Distribution by Normal Distribution

(6)

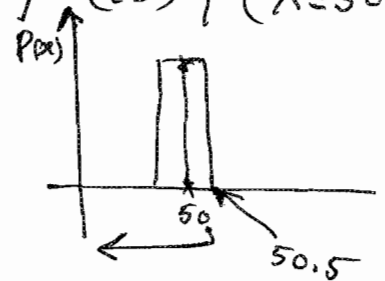
- Binomial probabilities can be obtained for  $n=1, 2, \dots, 10, 15, 20, 25$ . For this, we need approximation method for computing the binomial prob's.
- For large  $n$ , a normal distribution can be used to provide an approximation to binomial distribution.
- Note the binomial distribution is discrete dist. normal distribution is continuous dist.

They are different families

- If  $X \sim B(100, 0.4)$ , find (i)  $P(X \leq 50)$  (ii)  $P(X=50)$

(i)  $P_B(X \leq 50) \approx P_N(X \leq 50.5)$

[Continuity correction]

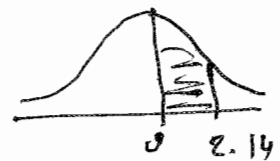


$$\mu = np = 100 \times 0.4 = 40$$

$$\sigma = \sqrt{100 \times 0.4 \times 0.6} = \sqrt{24} = 4.9$$

$$= P_N\left(Z \leq \frac{50.5 - 40}{4.9}\right)$$

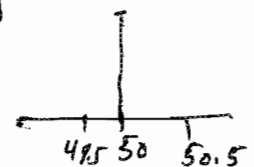
$$= P_N(Z \leq 2.14)$$



$$= \cancel{0.9838} T(2.14) = 0.9838$$

$$= \cancel{0.5 + 0.9838} \boxed{0.9838}$$

(ii)  $P_B(X=50) = P_N\left(\frac{49.5}{x_1} \leq X \leq \frac{50.5}{x_2}\right)$



$$x_1 \rightarrow z_1 = \frac{49.5 - 40}{4.9} = 1.94$$

$$x_2 \rightarrow z_2 = \frac{50.5 - 40}{4.9} = 2.14$$

$$P(1.94 < Z < 2.14) = T(2.14) - T(1.94) = 0.9838 - 0.9738 = 0.0100$$

Example 2: It is known that 60% of students are smoking, 7

A sample of size 50 students is taken. Find the probability that at least 40 students are smoking.

-  $X$ : # of smokers  $\sim B(50, 0.6)$

$$P_B(X \geq 40) = P(40) + P(41) + \dots + P(50)$$

You can't use Binomial Table

- We use normal approx.

$$\approx P_N(X \geq 39.5)$$

$$\mu = 50 \times 0.6 = \boxed{30}$$

$$\sigma = \sqrt{50 \times 0.6 \times 0.4} = \sqrt{12} = \boxed{3.46}$$

$$39.5 \longrightarrow z = \frac{39.5 - 30}{3.46} = \underline{2.75}$$

$$P(Z \geq 2.75) = 1.0 - T(2.75)$$

$$= 1.0 - 0.9970$$

$$= \boxed{0.003}$$

