

Summary on normal distribution

1. The Normal Distribution

- (a) The normal distribution with mean μ and variance σ^2 : $X \sim N(\mu, \sigma^2)$
- (b) The standard normal distribution: $\frac{X-\mu}{\sigma} = Z \sim N(0, 1)$
- i. The distribution curve is bell-shape.
 - ii. Use the standard normal table of $P(Z < z)$
 - iii. Z is symmetric about 0: $P(Z < z) = 1 - P(Z < -z)$
 - iv. $P(Z < -z) = P(Z > z)$
 - v. Standard normal percentiles and critical values:

Percentile	90	95	97.5	99	99.5
α (tail area)	.1	.05	.025	.01	.005
z_α	1.28	1.645	1.96	2.33	2.58

- vi. Values of $z_{\alpha/2}$:

$1 - \alpha$.80	.90	.95	.98	.99
$z_{\alpha/2}$	1.28	1.645	1.96	2.33	2.58

2. The Normal Approximation to the Binomial Distribution

If $X \sim \text{Bin}(n, p)$ with large n , then $Z = \frac{X-np}{\sqrt{np(1-p)}}$ is approximately standard normal.

(a) Type of the Problems

A discrete random variable X is involved. $X \sim \text{Bin}(n, p)$. Probability is required. Typically n is large, $np \geq 5$, $n(1-p) \geq 5$: approximation is appropriate.

(b) Method of approximation

Step 1 $\mu = np$, $\sigma^2 = np(1-p)$

Step 2 Use $N(\mu, \sigma^2) = N(np, np(1-p))$

Step 3 Remember to use continuity correction.

- i. $P(X = 3) = P\left(\frac{2.5-\mu}{\sigma} \leq Z \leq \frac{3.5-\mu}{\sigma}\right)$
- ii. $P(X < 3) = P\left(Z \leq \frac{2.5-\mu}{\sigma}\right)$
- iii. $P(X \leq 3) = P\left(Z \leq \frac{3.5-\mu}{\sigma}\right)$
- iv. $P(X > 3) = P\left(Z \geq \frac{3.5-\mu}{\sigma}\right)$
- v. $P(X \geq 3) = P\left(Z \geq \frac{2.5-\mu}{\sigma}\right)$

Example: $P(0 < X \leq 3) = P\left(\frac{.5-\mu}{\sigma} \leq Z \leq \frac{3.5-\mu}{\sigma}\right)$

3. Sampling Distributions

- (a) The r.v.'s X_1, X_2, \dots, X_n are a random sample (iid) of size n if
- X_1, X_2, \dots, X_n are independent.
 - Every X_i has the same probability distribution.
- (b) Let X_1, X_2, \dots, X_n be a random sample from a distribution with mean μ and standard deviation σ , then
- $E(\bar{X}) = \mu$
 - $\text{Var}(\bar{X}) = \sigma^2/n$ and $\sigma_{\bar{X}} = \sigma/\sqrt{n}$
- (c) Let X_1, X_2, \dots, X_n be a random sample from any distribution with mean μ and variance σ^2 . Then for large n ($n \geq 30$),
- \bar{X} is approximately $N(\mu, \sigma^2/n)$
 - Central Limit Theorem (CLT):

$$Z = \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \text{ is approximately } N(0, 1)$$

- (d) For unknown σ

If X_1, \dots, X_n is a random sample from $N(\mu, \sigma^2)$, and $S^2 = \sum_{i=1}^n (X_i - \bar{X})^2 / (n - 1)$, then

$$T = \frac{\bar{X} - \mu}{S/\sqrt{n}} \sim t_{n-1}$$

Example

Review on Sampling Distribution of the Mean

- Type of the Problems

(Population) mean μ and standard deviation σ of a random variable X is given. Then it asks for probability concerning the sample mean \bar{X} . (In some problems, probability of sum is required).

- II. Method

A. If $X \sim N(\mu, \sigma^2)$, $\bar{X} \sim N(\mu, \sigma^2/n)$, $n \geq 1$

B. If X is not normal, $\bar{X} \simeq N(\mu, \sigma^2/n)$, $n \geq 30$

- In any case, we calculate probability using $N(\mu, \sigma^2/n)$.

Do not use continuity correction.

Eg. X has mean $\mu = 5$, standard deviation $\sigma = 20$, $n = 100$.

\bar{X} is approximately $N(\mu, \sigma^2/n) = N(5, 4)$.

Hence $P(\bar{X} > 6) = P(Z > \frac{6-5}{2}) = P(Z > .5) = .3085$.