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Normal Distribution

A continuous random variable X is said to have normal distribution with parameters μ and σ if its probability density function is given by

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

where $-\infty < \mu < \infty$ and $\sigma > 0$.

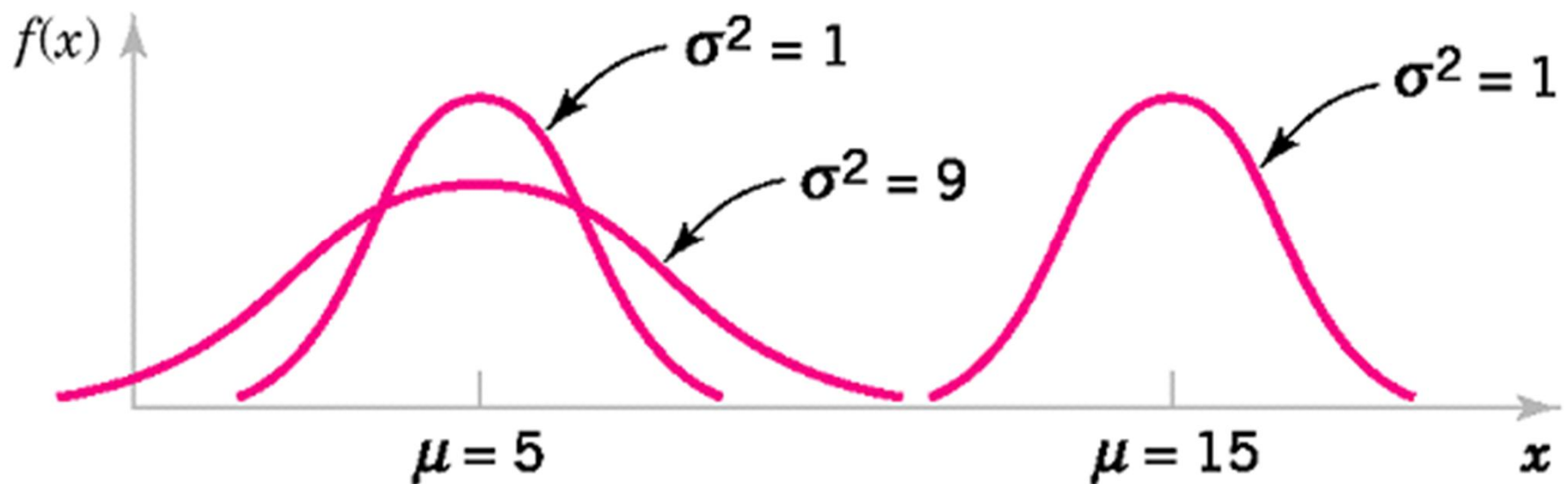


Normal Distribution

The normal distribution is also known as Gaussian distribution because the Gaussian function which is defined as

$$f(x) = a e^{-(x-b)^2 / c^2}$$

Normal Distribution



The parameter μ is called shape parameter and σ^2 is called scale parameter



Normal Distribution

Mean of Normal distribution:

$$\begin{aligned} E(X) &= \int_{-\infty}^{\infty} xf(x)dx = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \frac{x}{\sigma} e^{-(x-\mu)^2/2\sigma^2} dx \\ &= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \left(\frac{x-\mu}{\sigma} + \frac{\mu}{\sigma} \right) e^{-(x-\mu)^2/2\sigma^2} dx \\ &= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \sigma ue^{-u^2/2} du + \mu \int_{-\infty}^{\infty} f(x)dx. \\ &= 0 + \mu = \mu. \end{aligned}$$



Normal Distribution

Variance of Normal distribution:

$$\begin{aligned} \text{Var}(X) &= \int_{-\infty}^{\infty} (x - \mu)^2 \frac{1}{\sqrt{2\pi}\sigma} e^{-(x-\mu)^2/2\sigma^2} dx \\ &= \frac{\sigma^2}{\sqrt{2\pi}} \int_{-\infty}^{\infty} u^2 e^{-u^2/2} du, \text{ put } u = \frac{x - \mu}{\sigma} \\ &= \frac{2\sigma^2}{\sqrt{2\pi}} \int_0^{\infty} u^2 e^{-u^2/2} du \end{aligned}$$



Normal Distribution

Variance of Normal distribution:

$$\begin{aligned} \text{Var}(X) &= \frac{2\sigma^2}{\sqrt{2\pi}} \int_0^{\infty} u^2 e^{-u^2/2} du \\ &= \frac{2\sigma^2}{\sqrt{2\pi}} \left(\left[u \int u e^{-u^2/2} du \right]_0^{\infty} - \int_0^{\infty} 1 \cdot \left(\int u e^{-u^2/2} du \right) du \right) \\ &= \frac{2\sigma^2}{\sqrt{2\pi}} \left(\left[-u e^{-u^2/2} \right]_0^{\infty} + \int_0^{\infty} e^{-u^2/2} du \right) \\ &= 0 + \sigma^2. \end{aligned}$$



Normal Distribution

Moment generating function of Normal distribution:

$$\begin{aligned}M_X(t) &= E\left(e^{tX}\right) = \int_{-\infty}^{\infty} e^{tx} \frac{1}{\sqrt{2\pi\sigma}} e^{-(x-\mu)^2/2\sigma^2} dx \\&= \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi\sigma}} e^{-\frac{1}{2\sigma^2}[(x-\mu)^2 - 2\sigma^2 tx]} dx \\&= \frac{1}{\sqrt{2\pi\sigma}} \int_{-\infty}^{\infty} e^{-\frac{1}{2\sigma^2}[(x-\mu)^2 - 2\sigma^2 t(x-\mu) + \sigma^4 t^2 - \sigma^4 t^2 - 2\sigma^2 \mu t]} dx\end{aligned}$$



Normal Distribution

Moment generating function of Normal distribution(cont.):

$$= e^{\frac{\sigma^2 t^2}{2} + \mu t} \frac{1}{\sqrt{2\pi}\sigma} \int_{-\infty}^{\infty} e^{-\frac{1}{2\sigma^2}[(x-\mu)^2 - 2\sigma^2 t(x-\mu) + \sigma^4 t^2]} dx$$

$$= e^{\frac{\sigma^2 t^2}{2} + \mu t} \frac{1}{\sqrt{2\pi}\sigma} \int_{-\infty}^{\infty} e^{-\frac{1}{2\sigma^2}[(x-\mu) - \sigma^2 t]^2} dx$$

$$= e^{\frac{\sigma^2 t^2}{2} + \mu t}$$



Normal Distribution

Mean, variance and moment generating function of the Normal distribution:

$$E(X) = \mu$$

$$Var(X) = \sigma^2$$

$$M_X(t) = e^{\mu t + \frac{\sigma^2 t^2}{2}}$$



Standard Normal Distribution

Normal Distribution (Special Case):

A normal random variable with mean 0 and variance 1 is called **standard normal random variable**.

Note: If X is normal with mean μ and variance σ^2 , then $\frac{X - \mu}{\sigma}$ is standard normal distribution.



Standard Normal Distribution

If Z is a standard normal random variable, then the density function of Z is given by

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



Advantage of standard normal distribution

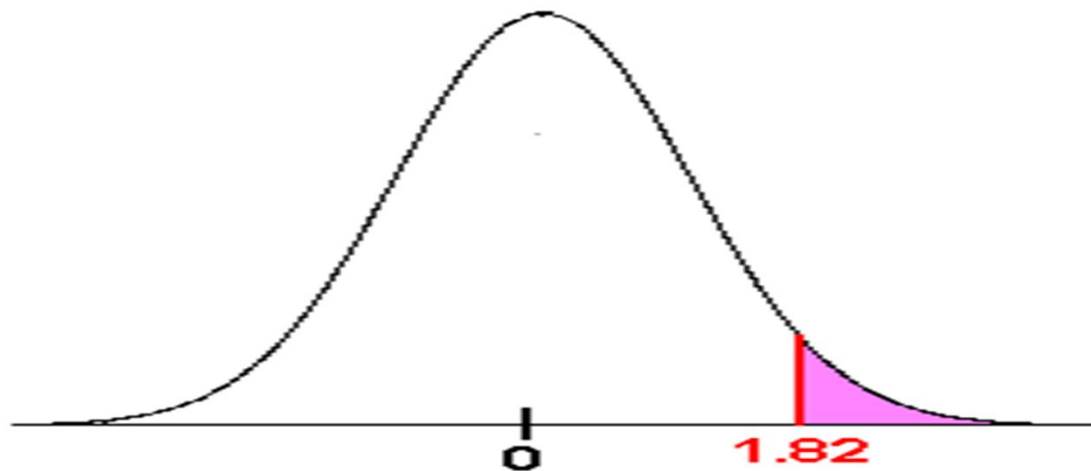
Remark:

- Cumulative distribution values available.
- Every distribution (discrete/continuous) can be approximated by normal distribution and hence standard normal distribution (the proof required central limit theorem which will be discussed in Chapter 7).

Standard normal distribution

Notation:

$$P(Z \geq z_{\alpha}) = \alpha$$



$$z_{0.0344} = 1.82 \text{ since } P(Z > 1.82) = 0.0344.$$

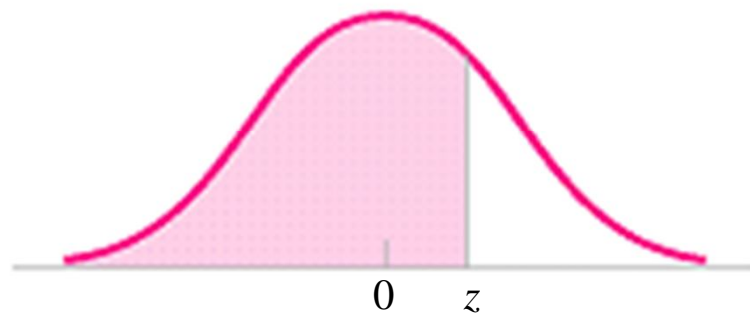
Standard normal distribution

How to use cumulative distribution Table:

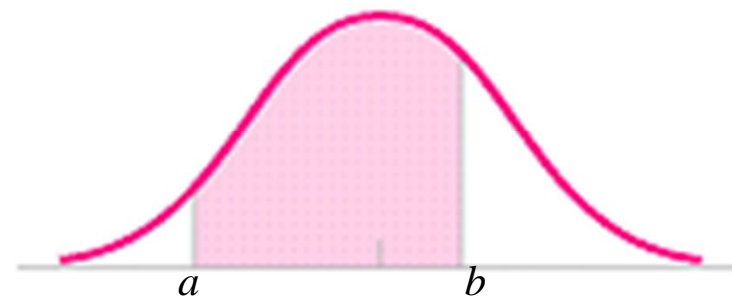
TABLE V
Cumulative distribution: Standard normal (*concluded*)

<i>z</i>	0.00	0.01	0.02	0.03	0.04	0.05	0.06
0.0	0.5000	0.5040	0.5080	0.5120 *	0.5160	0.5199	0.5239
0.1	0.5398	0.5438	0.5478	0.5517	0.5557	0.5596	0.5636
0.2	0.5793	0.5832	0.5871	0.5910	0.5948	0.5987	0.6026
0.3	0.6179	0.6217	0.6255	0.6293	0.6331	0.6368	0.6406
0.4	0.6554	0.6591	0.6628	0.6664	0.6700	0.6736	0.6772
0.5	0.6915	0.6950	0.6985	0.7019	0.7054	0.7088	0.7122
0.6	0.7257	0.7291	0.7324	0.7357	0.7389	0.7422	0.7454
0.7	0.7580	0.7611	0.7642	0.7673	0.7704	0.7734	0.7764
0.8	0.7881	0.7910	0.7939	0.7967	0.7995	0.8023	0.8051
0.9	0.8159	0.8186	0.8212	0.8238	0.8264	0.8289	0.8315
1.0	0.8413	0.8438	0.8461	0.8485	0.8508	0.8531	0.8554
1.1	0.8643	0.8665	0.8686	0.8708	0.8729	0.8749	0.8769
1.2	0.8849	0.8869	0.8888	0.8907	0.8925	0.8944	0.8962
1.3	0.9032	0.9049	0.9066	0.9082	0.9099	0.9115	0.9131

Standard normal Distribution



The standard normal probabilities
 $F(z) = P(Z \leq z)$



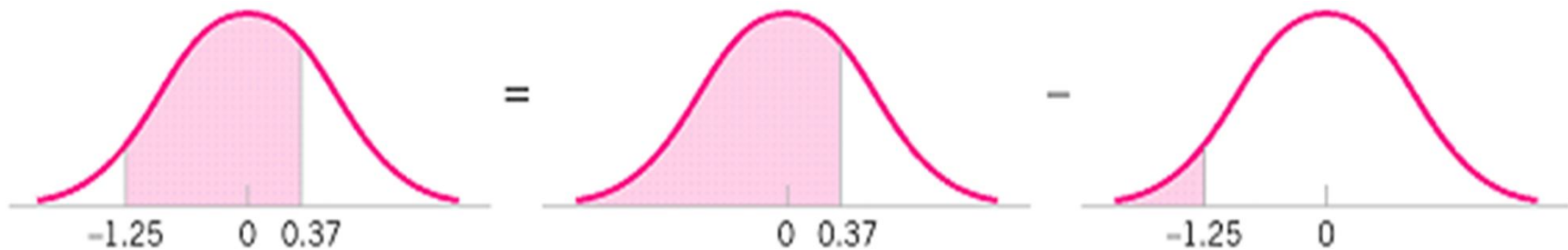
The standard normal probability
 $F(b) - F(a) = P(a \leq Z \leq b)$

Problem 1: $F(-z) = 1 - F(z)$ (verify yourself)

Standard normal distribution

Problem 2: Find the probabilities that a random variable having the standard normal distribution will take on a value (a) between -1.25 and 0.37 ; (b) greater than 1.26 ; (c) greater than -1.37 .

$$(a) P(-1.25 < Z < 0.37) = F(0.37) - F(-1.25) = 0.6443 - 0.1056 = 0.5387$$





Standard normal distribution

Problem 3: Find (a) $z_{0.01}$ (b) $z_{0.05}$

Since $F(z_{0.01}) = 1 - 0.01 = 0.99$, we look for the entry in Table which is closest to 0.99 and get 0.9901 corresponding to $z = 2.33$. Thus $z_{0.01} = 2.33$.

Similarly try part (b)



Normal to standard normal

If a random variable X has a normal distribution with the mean μ and the standard deviation σ , then, the probability that the random variable X will take on a value less than or equal to a , is given by

$$P(X \leq a) = P\left(\frac{X - \mu}{\sigma} \leq \frac{a - \mu}{\sigma}\right) = P\left(Z \leq \frac{a - \mu}{\sigma}\right) = F\left(\frac{a - \mu}{\sigma}\right)$$

Similarly

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



Standard normal distribution

Problem 4:

If a random variable has the normal distribution with $\mu = 16.2$ and $\sigma^2 = 1.5625$, find the probabilities that it will take on a value (a) greater than 16.8;

(b) less than 14.9; (c) between 13.6 and 18.8.

Solution:

$$\begin{aligned} \text{(a) } P(X > 16.8) &= 1 - P(X \leq 16.8) = 1 - F\left(\frac{16.8 - 16.2}{1.25}\right) \\ &= 1 - F(0.48) = 1 - 0.6844 = 0.3156 . \end{aligned}$$

$$\text{(b) } P(X < 14.9) = F\left(\frac{14.9 - 16.2}{1.25}\right) = F(-1.04) = 0.1492.$$

Problem 5



The average life of bulb is 1000 hours and the standard deviation is 300 hours. If X is the life period of a bulb which is distributed normally, find the probability that a randomly picked bulb will function for less than 500 hours.

Problem 6



Suppose that the load of an airplane wing is a random variable X with $N(1000; 14400)$ distribution. The maximum load that the wing can withstand is a random variable Y which is $N(1260; 2500)$. If X and Y are independent, find the probability that the load encountered by the wing is less than its critical load.

Problem 7



Let Z be a Standard Normal random variable.
Show that Z^2 follows Chi-Squared with
degree of freedom 1.

Problem 8



Suppose scores of 300 students in Statistics course has follows normal distribution. If an instructor want to assign A grades to those scores above $\mu + \alpha$ then how many students will get A grades ?