1 Why Is It Gaussian?

Let X be a normally distributed random variable with mean μ and variance σ^2 . Let Y = aX + b, where a > 0 and b are non-zero real numbers. Show explicitly that Y is normally distributed with mean $a\mu + b$ and variance $a^2\sigma^2$. The PDF for the Gaussian Distribution is $\frac{1}{\sqrt{2\pi\sigma^2}}e^{-\frac{(x-\mu)^2}{2\sigma^2}}$. One approach is to start with the cumulative distribution function of Y and use it to derive the probability density function of Y.

[1.You can use without proof that the pdf for any gaussian with mean and standard deviation is given by the formula $\frac{1}{\sqrt{2\pi\sigma^2}}e^{-\frac{(x-\mu)^2}{2\sigma^2}}$ where μ is the mean value for X and σ^2 is the variance. 2. The derivative of CDF gives PDF.]

$$\begin{array}{lll}
X \sim N(\mu, \sigma^2) & Y = aX + b & \text{for } a \neq 0 \\
F_{Y}(x) & = P(Y \leq x) \\
& = P(aX + b \leq x) \\
& = P(X \in \frac{x - b}{a}) & \text{since } a \neq 0 \\
& = F_{X}(\frac{x - b}{a}) & \exp(3) \neq e \\
\end{array}$$

$$\begin{array}{ll}
F_{Y}(x) & = \frac{d}{dx} F_{Y}(x) & g(x) = \frac{x - b}{a} & d f_{X}(g(x)) \\
& = \frac{d}{dx} F_{X}(\frac{x - b}{a}) & F_{X}(g(x)) & f_{X}'(g(x)) \\
& = \frac{1}{a} \frac{1}{1210^{2}} \exp\left(-\frac{(x - b - \mu)^{2}}{20^{2}}\right)
\end{array}$$

want
$$\left(\frac{x-b}{a}-\mu\right)^2 = \frac{\left(x-ca\mu+b\right)^2}{2a^2 \int_{a}^{2}}$$

$$\frac{x-b}{a}-\mu = \frac{1}{a}\left(x-b-a\mu\right)$$

$$\frac{1}{a\sqrt{2\pi}} \exp\left(-\frac{(x-b-a\mu)^2}{2a^2 \int_{a}^{2}}\right)$$

fy(x) is the pdf of gaussian r.v. w/

Central limit theorem

For $X_{1,1...}, X_{n}$ i.i.d. r.v.S W rean fe & Variance $\int_{1}^{2} \frac{S_{n}}{n} = \frac{1}{n} \frac{2}{n} X_{i}$ converges in distribution to $N(\mu, \frac{1}{n})$. $\#S_{n} = \frac{2}{n} X_{i}$

$$\frac{Sn}{n-\mu} = \frac{Sn-n\mu}{\sqrt{nc^2}} \times N(0,1)$$

Standard Normal - N(0,1)

$$f_{x}(x): \frac{1}{\sqrt{2\pi}} e^{-x^{2}/2}$$

2 Hypothesis testing

We would like to test the hypothesis claiming that a coin is fair, i.e. P(H) = P(T) = 0.5. To do this, we flip the coin n = 100 times. Let Y be the number of heads in n = 100 flips of the coin. We decide to reject the hypothesis if we observe that the number of heads is less than 50 - c or larger than 50 + c. However, we would like to avoid rejecting the hypothesis if it is true; we want to keep the probability of doing so less than 0.05. Please determine c. (Hints: use the central limit theorem to estimate the probability of rejecting the hypothesis given it is actually true. Table is provided in the appendix.)

if the hypothesis were true,
$$\mu = E(X_i)^2 = \frac{1}{2}$$
 $6^2 = Var(X_i) = \frac{1}{4}$

$$P(\frac{Y-nM}{\sqrt{n\sigma^2}} \in Z) \approx \phi(Z)$$
 since $\frac{Y-nM}{\sqrt{n\sigma^2}} \approx N(0,1)$

we reject the hypothesis when 14-50|7C want P(14-50|7C)<0.05 want P(14-50|4C)70.95

$$P(14-501 \le C) = P(\frac{14-501}{5} \le \frac{C}{5})$$
 $= P(-\frac{C}{5} \le \frac{4-50}{5} \le \frac{C}{5})$
 $\approx P(-\frac{C}{5} \le \frac{4-50}{5} \le \frac{C}{5})$

= $\phi(\xi) - (1 - \phi(\xi))$ = $2\phi(\xi) - 1$

$$2 \phi(\frac{c}{5}) - 1 = 0.95$$

$$\phi(\frac{c}{5}) = 0.975$$

$$c = \phi^{-1}(0.975)$$

$$c = 1.96$$

$$c = 9.8$$

if we see at least 50+10=60 reals or at most 50-10=40 reals, reject the hypothesis.

X~N(0,1) P(X < 1.86) = 0.9686

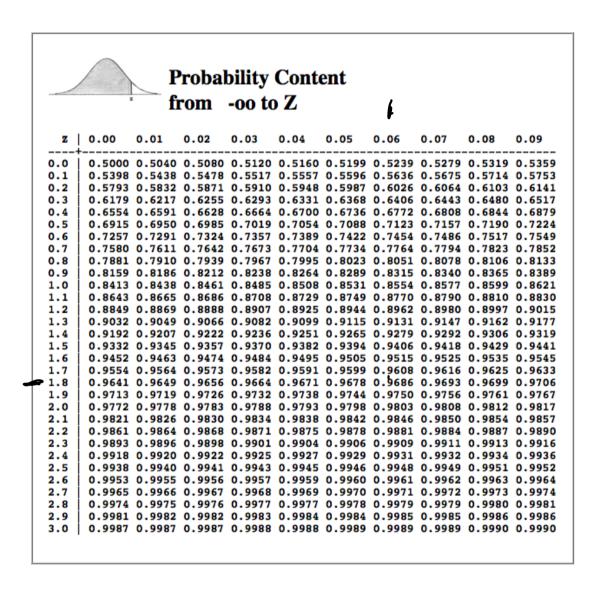


Table 1: Table of the Normal Distribution