

## Class 6 – Statistics and Probability (part II)

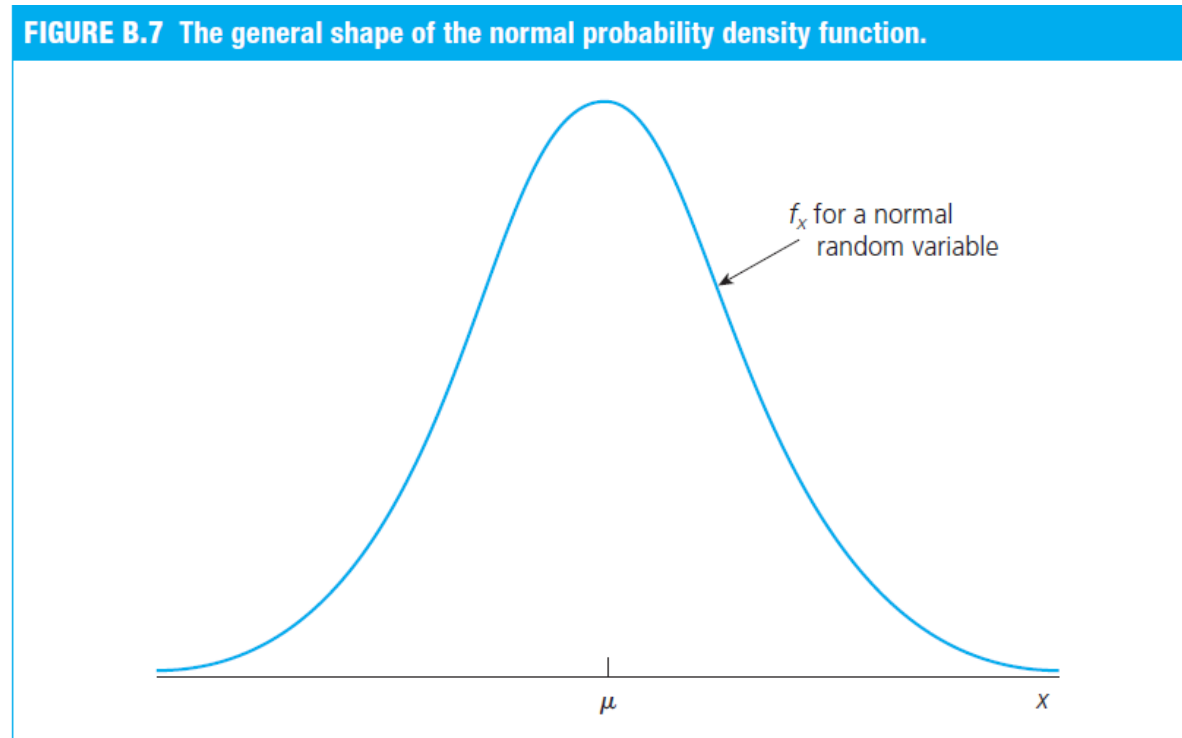
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# Normal distribution

A normal random variable is a continuous random variable that can take on any value. Its probability density function has the familiar bell shape graphed in Figure below.

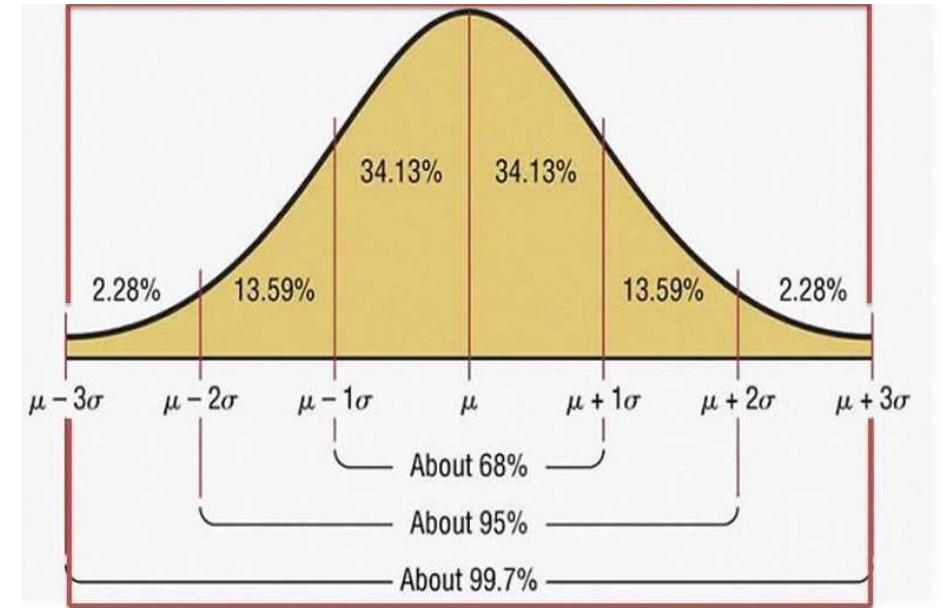


- We say that  $X$  has a normal distribution with expected value  $\mu$  and variance  $\sigma^2$ , written as  $X \sim N(\mu, \sigma^2)$
- Because the normal distribution is **symmetric**, mean, median and mode are the same.

# Properties of Normal distribution

## 1. Standard Normal Distribution

If  $X \sim N(\mu, \sigma^2)$  then  $\frac{X-\mu}{\sigma} \sim N(0, 1)$



2. If  $X \sim N(\mu, \sigma^2)$  then  $aX+b \sim N(a\mu + b, a^2\sigma^2)$

3. Any **linear combination** of iid normal random variables has a normal distribution!

We see this property again when we talk about the distribution of sample averages!

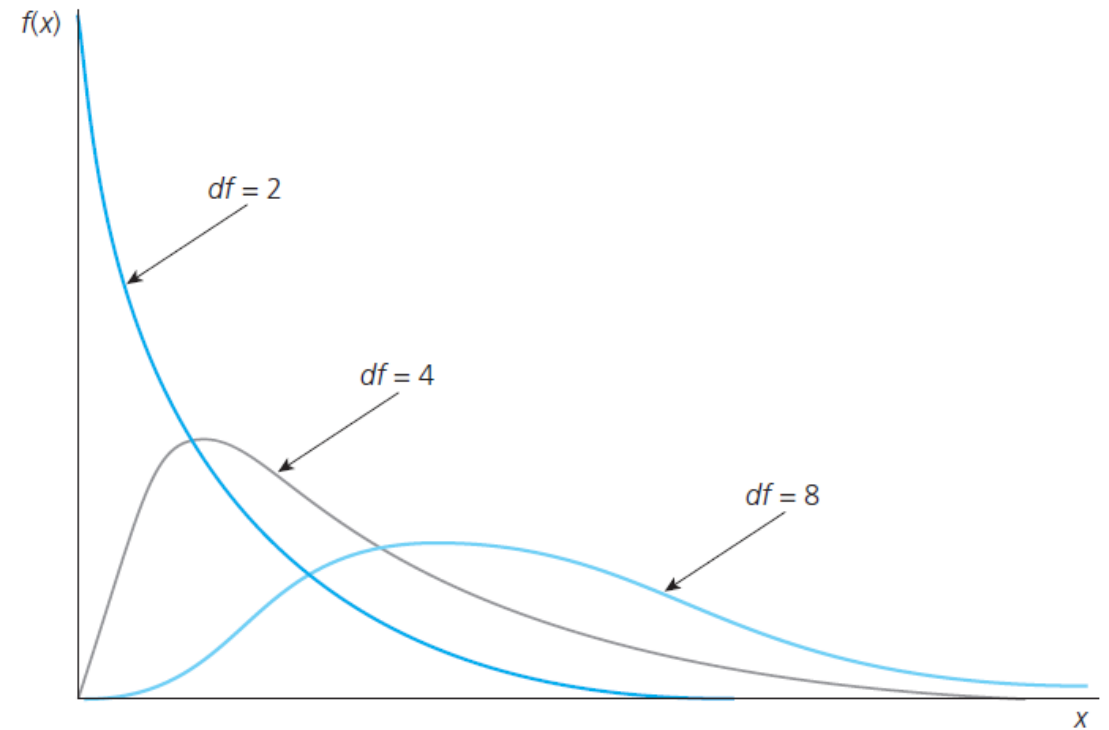
# The Chi-square distribution

The chi-square distribution is obtained directly from independent, standard normal random variables.

$$X = \sum_{i=1}^n Z_i^2$$

- ✓  $X$  has a chi-square distribution with  $n$  degrees of freedom (df).  $X \sim \chi_n^2$
- ✓ The df in a chi-square distribution corresponds to the number of terms in the sum
- ✓ chi-square random variable is always nonnegative and not symmetric

FIGURE B.9 The chi-square distribution with various degrees of freedom.



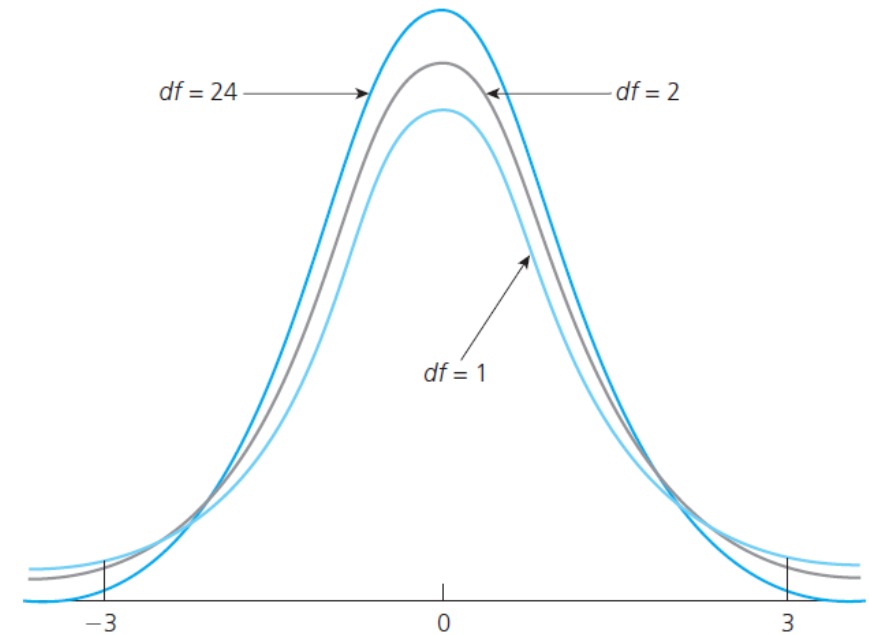
# The t distribution

The t distribution is the workhorse in classical statistics and multiple regression analysis. It is obtained from a standard normal and a chi-square random variable with  $n$  degrees of freedom.

$$T = \frac{Z}{\sqrt{X/n}}$$

- ✓  $T$  has a t distribution with  $n$  degrees of freedom.  $T \sim t_n$
- ✓ The t distribution gets its degrees of freedom from the  $\chi^2$  random variable
- ✓ The pdf of the t distribution has a shape similar to that of the standard normal distribution, except that it is more spread out and therefore has more area in the tails

FIGURE B.10 The  $t$  distribution with various degrees of freedom.



# The F distribution

The F distribution will be used for testing hypotheses in the context of multiple regression analysis. It is obtained from two independent **chi-square** random variables with  $k_1$ , and  $k_2$  degrees of freedom.

$$F = \frac{\left(\frac{X_1}{k_1}\right)}{\left(\frac{X_2}{k_2}\right)}$$

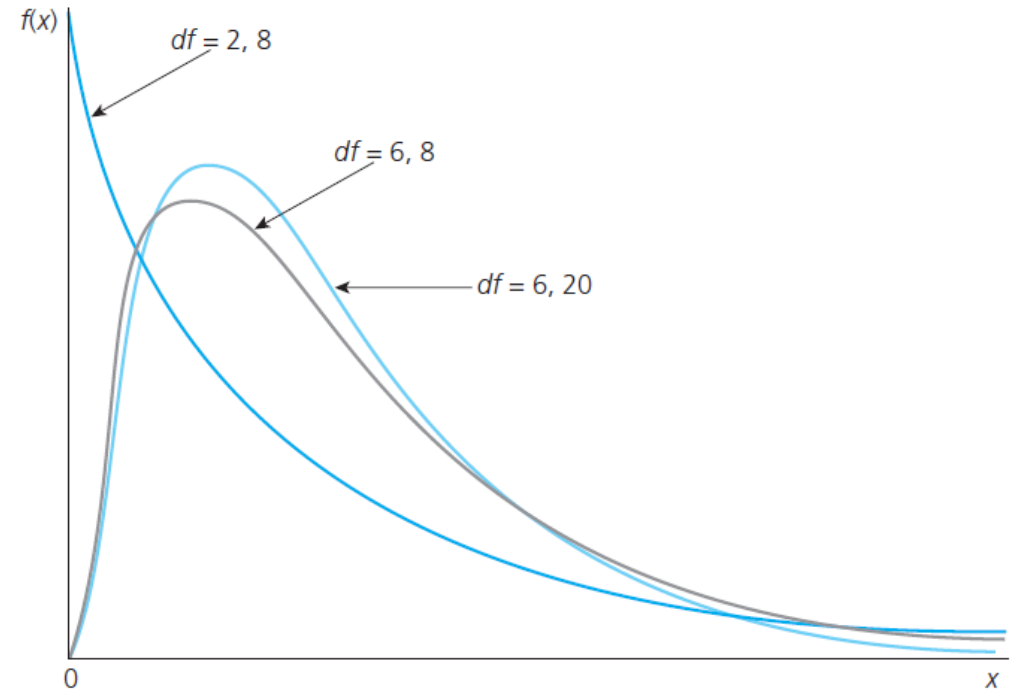
✓ F has a F-distribution with  $(k_1, k_2)$  degrees of freedom.

$$F \sim F(k_1, k_2)$$

✓  $k_1$  is called the **numerator** degrees of freedom

✓  $k_2$  is called the **denominator** degrees of freedom

FIGURE B.11 The  $F_{k_1, k_2}$  distribution for various degrees of freedom,  $k_1$  and  $k_2$ .



# Functions for descriptive statistics

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<b>mean(x)</b>	Sample average $\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i$
<b>median(x)</b>	sample median
<b>var(x)</b>	Sample variance $s_x^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$
<b>sd(x)</b>	Sample standard deviation $s_x = \sqrt{s_x^2}$
<b>cov(x, y)</b>	Sample covariance $c_{xy} = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})$
<b>cor(x, y)</b>	Sample correlation $r_{xy} = \frac{s_{xy}}{s_x \cdot s_y}$
<b>quantile(x, q)</b>	$q$ quantile = $100 \cdot q$ percentile, e.g. <b>quantile(x, 0.5)</b> = sample median

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# Functions for statistical distributions

Distribution	Param.	pmf/pdf	cdf	Quantile	Random numbers
<i>Discrete distributions:</i>					
Bernoulli	$p$	<b>dbinom</b> ( $x, 1, p$ )	<b>pbinom</b> ( $x, 1, p$ )	<b>qbinom</b> ( $q, 1, p$ )	<b>rbinom</b> ( $R, 1, p$ )
Binomial	$n, p$	<b>dbinom</b> ( $x, n, p$ )	<b>pbinom</b> ( $x, n, p$ )	<b>qbinom</b> ( $q, n, p$ )	<b>rbinom</b> ( $R, n, p$ )
Hypergeom.	$S, W, n$	<b>dhyper</b> ( $x, S, W, n$ )	<b>phyper</b> ( $x, S, W, n$ )	<b>qhyper</b> ( $q, S, W, n$ )	<b>rhyper</b> ( $R, S, W, n$ )
Poisson	$\lambda$	<b>dpois</b> ( $x, \lambda$ )	<b>ppois</b> ( $x, \lambda$ )	<b>qpois</b> ( $q, \lambda$ )	<b>rpois</b> ( $R, \lambda$ )
Geometric	$p$	<b>dgeom</b> ( $x, p$ )	<b>pgeom</b> ( $x, p$ )	<b>qgeom</b> ( $q, p$ )	<b>rgeom</b> ( $R, p$ )
<i>Continuous distributions:</i>					
Uniform	$a, b$	<b>dunif</b> ( $x, a, b$ )	<b>punif</b> ( $x, a, b$ )	<b>qunif</b> ( $q, a, b$ )	<b>runif</b> ( $R, a, b$ )
Logistic	—	<b>dlogis</b> ( $x$ )	<b>plogis</b> ( $x$ )	<b>qlogis</b> ( $q$ )	<b>rlogis</b> ( $R$ )
Exponential	$\lambda$	<b>dexp</b> ( $x, \lambda$ )	<b>pexp</b> ( $x, \lambda$ )	<b>qexp</b> ( $q, \lambda$ )	<b>rexp</b> ( $R, \lambda$ )
Std. normal	—	<b>dnorm</b> ( $x$ )	<b>pnorm</b> ( $x$ )	<b>qnorm</b> ( $q$ )	<b>rnorm</b> ( $R$ )
Normal	$\mu, \sigma$	<b>dnorm</b> ( $x, \mu, \sigma$ )	<b>pnorm</b> ( $x, \mu, \sigma$ )	<b>qnorm</b> ( $q, \mu, \sigma$ )	<b>rnorm</b> ( $R, \mu, \sigma$ )
Lognormal	$m, s$	<b>dlnorm</b> ( $x, m, s$ )	<b>plnorm</b> ( $x, m, s$ )	<b>qlnorm</b> ( $q, m, s$ )	<b>rlnorm</b> ( $R, m, s$ )
$\chi^2$	$n$	<b>dchisq</b> ( $x, n$ )	<b>pchisq</b> ( $x, n$ )	<b>qchisq</b> ( $q, n$ )	<b>rchisq</b> ( $R, n$ )
$t$	$n$	<b>dt</b> ( $x, n$ )	<b>pt</b> ( $x, n$ )	<b>qt</b> ( $q, n$ )	<b>rt</b> ( $R, n$ )
$F$	$m, n$	<b>df</b> ( $x, m, n$ )	<b>pf</b> ( $x, m, n$ )	<b>qf</b> ( $q, m, n$ )	<b>rf</b> ( $R, m, n$ )