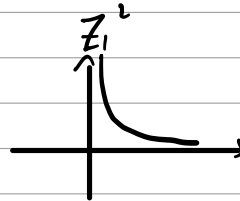
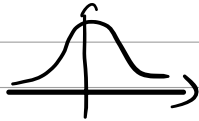


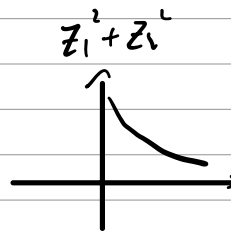
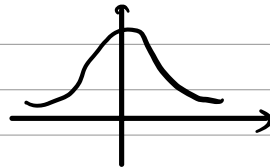
# 1. $\chi^2$ Chi-Square Distribution

pdf

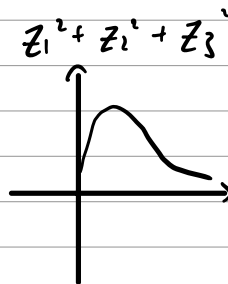
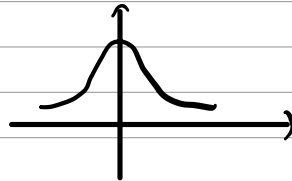
$$Z_1 \sim N(0, 1)$$



$$Z_2 \sim N(0, 1)$$



$$Z_3 \sim N(0, 1)$$



$$Y = \chi^2(k) = \sum_{i=1}^k Z_i^2$$

$$E[Y] = k \quad \text{Why? } E[Y] = \sum E[Z_i^2] = k$$

$$\text{Var}(Y) = 2k \quad \text{Why? } \text{Var}(Y) = \sum \text{Var}(Z_i^2)$$

$$= \sum \left[ \underbrace{E[Z_i^4]}_{\text{kurtosis}} - \underbrace{E[Z_i^2]^2}_1 \right]$$

= 3

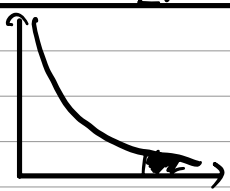
## 2. Goodness of Fit

	theory	sample
Left hand	9	11
Right hand	66	64

Test statistic:  $\chi^2 = \sum_{i=1}^k \frac{(O_i - e_i)^2}{e_i}$ ,  $df = 1$

$$\chi^2(1) = \frac{(11-9)^2}{9} + \frac{(64-66)^2}{66} = 0.505$$

差的越多, 值越大

Critical Value	p-value
 <p>5% level of significance</p>	<p>5% level of significance  <math>\Rightarrow p = 0.05</math>  or 5%</p>
<p><math>g(0) = 95\%</math>  <math>\alpha : g^{-1}(95\%) = 3.841</math></p>	<p><math>g(0.505) = 47.7\%</math></p>
<p><math>\therefore 0.505 &lt; 3.841,</math></p>	<p><math>\therefore</math> Not reject.</p>
<p><math>\therefore</math> Not reject.</p>	

3. Test for independence

Graf

	obs	expected
A	22	33.3
B	19	33.3
C	59	33.3

$$\chi^2(2) = \frac{(22 - 33.3)^2}{33.3} + \dots$$

	X	Y	total
A	10	12	22
B	9	10	19
C	25	34	59
	<u>44</u>	<u>56</u>	<u>100</u>

Q: (A, B, C) indep of (X, Y) ?

A:  $P(E1 \cap E2) = P(E1) \cdot P(E2)$

If indep, what to expect ?

A	22%
B	19%
C	59%

X	Y
44%	56%

G.F

	obs	expected
A	22	33.3
B	19	33.3
C	59	33.3

Indep

	obs (X)	expected
A	10	$22\% \times 44\% \div 100 = 9.68$
B	9	$19\% \times 44\% \div 100 = 8.36$
C	25	$59\% \times 56\% \div 100 = 33.04$

$$\chi^2_{(2)} = \frac{(10 - 9.68)^2}{9.68} + \frac{(9 - 8.36)^2}{8.36} + \frac{(25 - 33.04)^2}{33.04}$$

In general,

x y z

A

B

C

1)

$$df = (c-1)(r-1)$$

$$= 2 \times 3$$

$$= 6$$

4. Where is Normal distribution?

$$\chi^2_{(1)} = \underbrace{z_1^2}_{?}$$

CLT !

$$\left. \begin{array}{l} \text{Left} \\ \text{Right} \end{array} \right\} X = \text{Bern}(p) \quad \mu = p \quad \sigma^2 = p(1-p)$$

$$\frac{\sum X_i}{n} \stackrel{\text{CLT}}{\sim} N\left(\mu, \frac{\sigma^2}{n}\right)$$

$$\sum X_i \sim N(np, np(1-p))$$

$$Y = \sum X_i \sim \text{Bin}(n, p)$$

$$Y \sim N(np, np(1-p))$$

$$\frac{Y - np}{\sqrt{np(1-p)}} \sim Z$$

$$\frac{(Y - np)^2}{np(1-p)} \sim Z^2 = \chi^2_{(1)}$$

recall:

	theory	sample
Left hand	9 $np$	11 $Y$
Right hand	66 $n(1-p)$	64 $n-Y$

$$\chi^2 = \sum_{i=1}^k \frac{(O_i - E_i)^2}{E_i}$$

$$= \frac{(Y - np)^2}{np} + \frac{(n - Y - n(1-p))^2}{n(1-p)}$$

$$\text{Given } \frac{(Y - np)^2}{np(1-p)} \sim \chi^2 = \chi^2(1)$$

$$= \frac{(Y - np)^2 (p + (1-p))}{np(1-p)}$$

$$= \frac{(Y - np)^2}{n(1-p)} + \frac{(Y - np)^2}{np}$$

$$= \frac{(Y - np + n - n)^2}{n(1-p)} + \frac{(Y - np)^2}{np}$$

$$= \frac{(Y - n + n(1-p))^2}{n(1-p)} + \frac{(Y - np)^2}{np}$$

$$= \frac{(n - Y - n(1-p))^2}{n(1-p)} + \frac{(Y - np)^2}{np} \quad (\checkmark)$$