

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(\mu, \sigma^2)$$

$$\rightarrow Z_1 = \frac{X_1 - \mu}{\sigma}, Z_2 = \frac{X_2 - \mu}{\sigma}, \dots, Z_n = \frac{X_n - \mu}{\sigma} \stackrel{\text{iid}}{\sim} N(0, 1)$$

$$\vec{Z}^T \vec{Z} = \sum Z_i^2 = \sum \left(\frac{X_i - \mu}{\sigma} \right)^2 = \text{algebra} = \frac{(n-1)S_n^2}{\sigma^2} + n \frac{(\bar{x} - \mu)^2}{\sigma^2} \sim \chi_n^2$$

Fact: $U_1 \sim \chi_{k_1}^2$ indep. of $U_2 \sim \chi_{k_2}^2 \rightarrow U + V \sim \chi_{k_1+k_2}^2$

Recall from MATH 241

$$\bar{X} \sim N\left(\mu, \frac{\sigma^2}{n}\right) \rightarrow \frac{\bar{X} - \mu}{\frac{\sigma}{\sqrt{n}}} \sim N(0, 1)$$

$$\rightarrow \frac{(\bar{X} - \mu)^2}{\frac{\sigma^2}{n}} \sim \chi_1^2$$

$$\rightarrow n \frac{(\bar{X} - \mu)^2}{\sigma^2} \sim \chi_1^2$$

Conjecture:

$$1) \frac{(n-1)S_n^2}{\sigma^2} \sim \chi_{n-1}^2$$

$$2) \frac{(n-1)S_n^2}{\sigma^2} \text{ indep of } n \frac{(\bar{X} - \mu)^2}{\sigma^2}$$

$\longleftrightarrow S_n^2$ indep of \bar{X}_n
Since n, μ, σ are constants

$$\vec{z}^T \vec{z} \sim \chi^2_n$$

$$= \underbrace{\vec{z}^T I_n \vec{z}}_{\text{quadratic}}$$

$$\vec{z}^T \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & & & \\ \vdots & & & \\ 0 & & & 0 \end{bmatrix} \vec{z} = z_1^2 \sim \chi^2_1$$

\uparrow
 B_1

$$\vec{z}^T \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & \ddots \\ & & & 0 \end{bmatrix} \vec{z} = z_2^2 \sim \chi^2_1$$

\uparrow
 B_2

$$\vec{z}^T \begin{bmatrix} 0 & \dots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \dots & 1 \end{bmatrix} \vec{z} = z_n^2 \sim \chi^2_1$$

All of these are indep.

$$\vec{z}^T B_1 \vec{z} + \dots + \vec{z}^T B_n \vec{z} \sim \chi^2_n$$

$$\vec{z}^T (B_1 \vec{z} + B_2 \vec{z} + \dots + B_n \vec{z})$$

$$\vec{z}^T \underbrace{(B_1 + B_2 + \dots + B_n)}_{I_n} \vec{z}$$

$$\text{rank}[B_1] = \text{rank}[B_2] = \dots = \text{rank}[B_n] = 1$$

$$\sum \text{rank}[B_i] = n$$

Cochran's Theorem

If (a) $B_1 + \dots + B_k = I_n$ and

(b) $\sum_{j=1}^k \text{rank}[B_j] = n$

then

(a) $\vec{z}^T B_j \vec{z} \sim \chi^2_{\text{rank}[B_j]}$

(b) $\vec{z}^T B_{j_1} \vec{z}$ indep of $\vec{z}^T B_{j_2} \vec{z}$ for all $j_1 \neq j_2$

$$\vec{z}^T \vec{z} = \sum z_i^2 = \sum ((z_i - \bar{z}) + \bar{z})^2$$

$$= \sum (z_i - \bar{z})^2 + 2 \sum (z_i - \bar{z}) \bar{z} + \sum \bar{z}^2$$

$$= \sum (z_i - \bar{z})^2 + 2 \left(\sum z_i \bar{z} - \sum \bar{z}^2 \right) + \sum \bar{z}^2$$

$$= \sum (z_i - \bar{z})^2 + 2(n\bar{z}^2 - n\bar{z}^2) + n\bar{z}^2$$

Let $\vec{1} := \begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix}$ dim = n

$$\bar{z} = \frac{1}{n} \sum z_i = \frac{1}{n} \vec{1}^T \vec{z} = \frac{1}{n} \vec{z}^T \vec{1}_n$$

$$n\bar{z}^2 = n \bar{z} \bar{z} = n \left(\frac{1}{n} \vec{z}^T \vec{1}_n \right) \left(\frac{1}{n} \vec{1}_n^T \vec{z} \right)$$

$$= \vec{z} \frac{1}{n} \vec{1} \vec{1}^T \vec{z}$$

$$\text{Let } J_n = \begin{matrix} \xrightarrow{1} & \xrightarrow{1} \\ \text{rank } 1 & \text{rank } 1 \end{matrix} = \begin{bmatrix} 1 & 1 & \dots & 1 \\ \vdots & \vdots & & \vdots \\ 1 & 1 & \dots & 1 \end{bmatrix}$$

$$\vec{z}^T \frac{1}{n} \vec{1} \vec{1}^T \vec{z} = \underbrace{\vec{z}^T \left(\frac{1}{n} J_n \right) \vec{z}}_{B_2}$$

$$B_2 = \begin{bmatrix} \frac{1}{n} & & & \\ & \ddots & & \\ & & \ddots & \\ & & & \frac{1}{n} \end{bmatrix} \quad \text{rank } [B_2] = 1$$

$$\sum (z_i - \bar{z})^2 = \sum z_i^2 - 2 \sum z_i \bar{z} + \sum \bar{z}^2 = \sum z_i^2 - n \bar{z}^2$$

$$= \vec{z}^T J_n \vec{z} - \vec{z}^T \left(\frac{1}{n} J_n \right) \vec{z} \cdot n = (\vec{z} - \bar{z} \vec{1})^T (\vec{z} - \bar{z} \vec{1})$$

$$= \vec{z}^T \left(J_n - \frac{1}{n} J_n \right) \vec{z}$$

$$B_1 = \begin{bmatrix} 1 - \frac{1}{n} & -\frac{1}{n} & \dots & -\frac{1}{n} \\ -\frac{1}{n} & 1 - \frac{1}{n} & & \\ \vdots & & \ddots & \\ -\frac{1}{n} & & & 1 - \frac{1}{n} \end{bmatrix}$$

Theorem If Square matrix A is symmetric and idempotent i.e. $AA = A$ then $\text{rank } [A] = \text{tr}[A]$

$$(J_n - \frac{1}{n} J_n)^T = J_n^T - \frac{1}{n} J_n^T = J_n - \frac{1}{n} J_n$$

$$(I_n - \frac{1}{n} J_n)(I_n - \frac{1}{n} J_n)$$

$$= I_n I_n - \frac{1}{n} J_n I_n - \frac{1}{n} I_n J_n + \frac{1}{n^2} J_n J_n$$

$$= I_n - \frac{2}{n} J_n + \frac{1}{n} \cdot n \cdot J_n$$

$$= I_n - \frac{1}{n} J_n$$

$$\text{So rank}[B_i] = \sum_{i=1}^n 1 - \frac{1}{n} = n - 1$$

By Cochran's Thm,

$$\sum (z_i - \bar{z})^2 \sim \chi^2_{n-1} \text{ indep of } n\bar{z}^2 \sim \chi^2_1$$

$$\bar{z} = \frac{z_1 + \dots + z_n}{n} = \frac{\frac{x_1 - \mu}{\sigma} + \dots + \frac{x_n - \mu}{\sigma}}{n}$$

$$= \frac{x_1 + \dots + x_n - n\mu}{n\sigma}$$

$$= \frac{\bar{x} - \mu}{\sigma}$$

$$\Rightarrow n\bar{z}^2 \Rightarrow \frac{n(\bar{x} - \mu)^2}{\sigma^2} \sim \chi^2_1$$

$$\sum (z_i - \bar{z})^2 = \sum \left(\frac{x_i - \mu}{\sigma} - \frac{\bar{x} - \mu}{\sigma} \right)^2$$

$$= \sum \frac{(x_i - \bar{x})^2}{\sigma^2}$$

$$= \frac{(n-1)S^2}{\sigma^2} \sim \chi_{n-1}^2 \text{ which indep of } \bar{X}$$

$\rightarrow S^2$ and \bar{X} are independent

Fisher proved this in 1925

Greeny (1936) showed that only the normal distribution has this property.

If σ is known and you're testing

$$H_0: \mu = \text{Value}$$

$$\frac{\bar{X} - \mu}{\frac{\sigma}{\sqrt{n}}} \sim N(0,1) \quad \text{"One-Sample Z-Test"}$$

If we wish to test

$$H_0: \sigma^2 = \text{value}$$

$$\frac{(n-1)S^2}{\sigma^2} \sim \chi_{n-1}^2 \quad \text{"One-Sample } \chi^2 \text{ Test"}$$

If we have two independent samples and wish to test

$$H_0: \sigma_1^2 = \sigma_2^2 \quad \frac{\frac{(n_1-1)S_1^2}{\sigma_1^2} / (n_1-1)}{\frac{(n_2-1)S_2^2}{\sigma_2^2} / (n_2-1)}$$

$$= \frac{S_1^2 / \sigma_1^2}{S_2^2 / \sigma_2^2} = \frac{S_1^2}{S_2^2} \sim F_{n_1, n_2} \quad \text{"Two Sample F-Test for Variance equality"}$$

$$\frac{\bar{X} - \mu}{S} = \frac{\bar{X} - \mu}{\sqrt{\frac{S^2}{n}}} = \frac{\bar{X} - \mu}{\frac{1}{\sqrt{n}} \sqrt{S^2}} = \frac{\bar{X} - \mu}{\frac{1}{\sqrt{n}} \sqrt{\frac{\sigma^2}{n-1} S^2}}$$

σ unknown

If testing $H_0: \mu = \text{Value}$

"One-Sample T-test"

$$= \frac{\bar{X} - \mu}{\frac{\sigma}{\sqrt{n}}} \sim N(0, 1)$$

$$\sqrt{\frac{(n-1)S^2}{\sigma^2} / (n-1)} \sim \sqrt{\frac{\chi_{n-1}^2}{n-1}}$$

And the numerator and denominator are independent. (By Cochran's Theorem)

Multivariate Normal Distribution

$$Z_1, \dots, Z_n \stackrel{\text{iid}}{\sim} N(0, 1)$$

$$\text{Let } \vec{Z} = \begin{bmatrix} Z_1 \\ \vdots \\ Z_n \end{bmatrix}, E[\vec{Z}] = \vec{0}_n$$

$$\text{Var}[\vec{Z}] = I_n$$

$$\vec{Z} \sim f_{\vec{Z}}(\vec{Z}) = ?$$

$$f_{Z_1, \dots, Z_n}(Z_1, \dots, Z_n) \stackrel{\text{iid}}{=} \prod_{i=1}^n f(Z_i)$$

$$= \prod_{i=1}^n \frac{1}{\sqrt{2\pi}} e^{-\frac{Z_i^2}{2}}$$

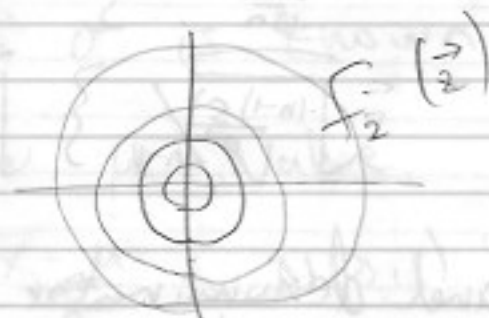
$$= \frac{1}{(2\pi)^{n/2}} e^{-\frac{1}{2} \sum z_i^2}$$

$$= \frac{1}{(2\pi)^{n/2}} e^{-\frac{1}{2} \vec{z}^T \vec{z}}$$

$$= N_n(\vec{0}, I_n), \text{ Standard Multivariate Normal Distribution}$$

\nwarrow mean
 \swarrow dimension \nearrow Variance

$$\vec{z} \sim N_2\left(\begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}\right)$$



Consider $\vec{\mu} \in \mathbb{R}^n$

$$\vec{x} = \vec{z} + \vec{\mu} = \begin{bmatrix} z_1 + \mu_1 \\ z_2 + \mu_2 \\ \vdots \\ z_n + \mu_n \end{bmatrix} \sim \begin{matrix} N(\mu_1, 1) \\ N(\mu_2, 1) \\ \vdots \\ N(\mu_n, 1) \end{matrix}$$

$$\vec{x} \sim N_n(\vec{\mu}, I_n)$$

$$\text{Let } A = \begin{bmatrix} 1 & 0 & 0 & \dots & 0 \\ 0 & 1 & 0 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & 1 \end{bmatrix}$$

$$\vec{x} = A\vec{z} = \begin{bmatrix} z_1 \\ z_1 + z_2 \\ z_1 + z_2 + z_3 \\ \vdots \\ z_1 + \dots + z_n \end{bmatrix} \begin{array}{l} \sim N(0,1) \\ \sim N(0,2) \\ \sim N(0,3) \\ \\ \sim N(0,n) \end{array} \left. \vphantom{\begin{bmatrix} z_1 \\ z_1 + z_2 \\ z_1 + z_2 + z_3 \\ \vdots \\ z_1 + \dots + z_n \end{bmatrix}} \right\} \begin{array}{l} \text{not} \\ \text{independent} \end{array}$$

$$\begin{aligned} \text{Cov}[z_1 + z_2, z_1] &= \text{Cov}[z_1, z_1] + \text{Cov}[z_1, z_2] \\ &= \text{Var}[z_1] + 0 \\ &= 1 \neq 0 \rightarrow \text{dependent} \end{aligned}$$

$$E[\vec{x}] = E[A\vec{z}], \quad \text{Var}[\vec{x}] = \text{Var}[A\vec{z}].$$