

L18 χ^2 -distributions in ANOVA tables

1. χ^2 -distributions in ANOVA table I

(1) χ^2 -distributions

Recall: $X \sim N(\mu, \Sigma)$, $A\Sigma A = A = A' \implies X'AX \sim \chi^2(\mu' A\mu, \text{tr}(A\Sigma))$.

(2) ANOVA table for M: $Y \sim N(X\beta, \sigma^2\Sigma)$ where $\mathcal{R}(1) \subset \mathcal{R}(X)$ and $\text{rank}(X) = r$.

Source	SS	DF
Model	$\text{SSM} = (\Sigma^{-1/2}Y)' \left[(\Sigma^{-1/2}X)(\Sigma^{-1/2}X)^+ - (\Sigma^{-1/2}1)(\Sigma^{-1/2}1)^+ \right] (\Sigma^{-1/2}Y)$	$r - 1$
Error	$\text{SSE} = (\Sigma^{-1/2}Y)' \left[I - (\Sigma^{-1/2}X)(\Sigma^{-1/2}X)^+ \right] (\Sigma^{-1/2}Y)$	$n - r$
C.Total	$\text{C.SSTO} = (\Sigma^{-1/2}Y)' \left[I - (\Sigma^{-1/2}1)(\Sigma^{-1/2}1)^+ \right] (\Sigma^{-1/2}Y)$	$n - 1$

(3) χ^2 -distributions in ANOVA table I

Note that there is H_0 under which model M is reduced to $M_1 : Y \sim N(1_n\gamma, \sigma^2\Sigma)$.

(i) $\frac{\text{SSE}}{\sigma^2} \sim \chi^2(n - r)$ (ii) $\frac{\text{SSM}}{\sigma^2} \stackrel{H_0}{\sim} \chi^2(r - 1)$ (iii) $\frac{\text{C.SSTO}}{n-1} \stackrel{H_0}{\sim} \chi^2(n - 1)$

Proof. (i) $Y \sim N(X\beta, \sigma^2\Sigma) \implies \Sigma^{-1/2}Y \sim N(\Sigma^{-1/2}X\beta, \sigma^2I_n)$.

$\frac{\text{SSE}}{\sigma^2} = (\Sigma^{-1/2}Y)' A (\Sigma^{-1/2}Y)$ where $A = \frac{I - (\Sigma^{-1/2}X)(\Sigma^{-1/2}X)^+}{\sigma^2}$.

But $A\sigma^2I_n A = A = A'$, $(\Sigma^{-1/2}X\beta)' A (\Sigma^{-1/2}X\beta) = 0$ and $\text{tr}(A\sigma^2I_n) = n - r$.

So $\frac{\text{SSE}}{\sigma^2} \sim \chi^2(n - r)$.

(ii) Under H_0 , $Y \sim N(1_n\gamma, \sigma^2\Sigma) \implies \Sigma^{-1/2}Y \sim N(\Sigma^{-1/2}1_n\gamma, \sigma^2I_n)$.

$\frac{\text{SSM}}{\sigma^2} = (\Sigma^{-1/2}Y)' A (\Sigma^{-1/2}Y)$ where $A = \frac{(\Sigma^{-1/2}X)(\Sigma^{-1/2}X)^+ - (\Sigma^{-1/2}1_n)(\Sigma^{-1/2}1_n)^+}{\sigma^2}$.

But $A\sigma^2I_n A = A = A'$, $(\Sigma^{-1/2}1_n\gamma)' A (\Sigma^{-1/2}1_n\gamma) = 0$, $\text{tr}(A\sigma^2I_n) = r - 1$.

So under H_0 , $\frac{\text{SSM}}{\sigma^2} \sim \chi^2(r - 1)$.

(iii) Under H_0 , $Y \sim N(1_n\gamma, \sigma^2\Sigma) \implies \Sigma^{-1/2}Y \sim N(\Sigma^{-1/2}1_n\gamma, \sigma^2I_n)$.

$\frac{\text{C.SSTO}}{\sigma^2} = (\Sigma^{-1/2}Y)' A (\Sigma^{-1/2}Y)$ where $A = \frac{I - (\Sigma^{-1/2}1_n)(\Sigma^{-1/2}1_n)^+}{\sigma^2}$.

But $A\sigma^2I_n A = A = A'$, $(\Sigma^{-1/2}1_n\gamma)' A (\Sigma^{-1/2}1_n\gamma) = 0$, $\text{tr}(A\sigma^2I_n) = n - 1$.

So under H_0 , $\frac{\text{C.SSTO}}{\sigma^2} \sim \chi^2(n - 1)$.

Ex1: For $Y \sim N(X\beta, \sigma^2I_n)$ with $\mathcal{R}(1_n) \subset \mathcal{R}(X)$, in ANOVA table

Source	SS	DF
Model	$\text{SSM} = Y' (XX^+ - 1_n 1_n^+) Y$	$r - 1$
Error	$\text{SSE} = Y' (I_n - XX^+) Y$	$n - r$
C.Total	$\text{C.SSTO} = Y' (I_n - 1_n 1_n^+) Y$	$n - 1$

$\frac{\text{SSM}}{\sigma^2} \stackrel{H_0}{\sim} \chi^2(r - 1)$, $\frac{\text{SSE}}{\sigma^2} \sim \chi^2(n - r)$ and $\frac{\text{C.SSTO}}{\sigma^2} \stackrel{H_0}{\sim} \chi^2(n - 1)$.

2. χ^2 -distributions in ANOVA table II

(1) ANOVA table for M: $Y \sim N(X\beta, \sigma^2\Sigma)$ where $\mathcal{R}(1) \not\subset \mathcal{R}(X)$ and $\text{rank}(X) = r$.

Source	SS	DF
Model	$\text{SSM} = (\Sigma^{-1/2}Y)' (\Sigma^{-1/2}X)(\Sigma^{-1/2}X)^+ (\Sigma^{-1/2}Y)$	r
Error	$\text{SSE} = (\Sigma^{-1/2}Y)' \left[I - (\Sigma^{-1/2}X)(\Sigma^{-1/2}X)^+ \right] (\Sigma^{-1/2}Y)$	$n - r$
U.Total	$\text{U.SSTO} = (\Sigma^{-1/2}Y)' I_n (\Sigma^{-1/2}Y)$	n

(2) χ^2 -distributions in ANOVA table II

Note that under $H_0 : \beta = 0$ model M is reduced to $M_0 : Y \sim N(0, \sigma^2\Sigma)$.

(i) $\frac{\text{SSE}}{\sigma^2} \sim \chi^2(n - r)$ (ii) $\frac{\text{SSM}}{\sigma^2} \stackrel{H_0}{\sim} \chi^2(r)$ (iii) $\frac{\text{U.SSTO}}{n} \stackrel{H_0}{\sim} \chi^2(n)$

Proof. (i) See 1 (3) (i).

(ii) Under $H_0 : \beta = 0, Y \sim N(0, \sigma^2 \Sigma) \implies \Sigma^{-1/2} Y \sim N(0, \sigma^2 I_n)$.

$$\frac{SSM}{\sigma^2} = (\Sigma^{-1/2} Y)' A (\Sigma^{-1/2} Y) \text{ where } A = \frac{(\Sigma^{-1/2} X)(\Sigma^{-1/2} X)^+}{\sigma^2}.$$

$$\text{But } A \sigma^2 I_n A = A = A', 0' A 0 = 0, \text{tr}(A \sigma^2 I_n) = r.$$

$$\text{So under } H_0, \frac{SSM}{\sigma^2} \sim \chi^2(r).$$

(iii) Under $H_0 : \beta = 0, Y \sim N(0, \sigma^2 \Sigma) \implies \Sigma^{-1/2} Y \sim N(0, \sigma^2 I_n)$.

$$\frac{U.SSTO}{\sigma^2} = (\Sigma^{-1/2} Y)' A (\Sigma^{-1/2} Y) \text{ where } A = \frac{I_n}{\sigma^2}.$$

$$\text{But } A \sigma^2 I_n A = A = A', 0' A 0 = 0, \text{tr}(A \sigma^2 I_n) = n.$$

$$\text{So under } H_0, \frac{U.SSTO}{\sigma^2} \sim \chi^2(n).$$

Ex2: For $Y \sim N(X\beta, \sigma^2 I_n)$ with $\mathcal{R}(1_n) \not\subset \mathcal{R}(X)$, in ANOVA table

Source	SS	DF
Model	$SSM = Y' X X^+ Y$	r
Error	$SSE = Y' (I_n - X X^+) Y$	$n - r$
U.Total	$U.SSTO = Y' I_n Y$	n

$$\frac{SSM}{\sigma^2} \stackrel{H_0}{\sim} \chi^2(r), \frac{SSE}{\sigma^2} \sim \chi^2(n - r) \text{ and } \frac{U.SSTO}{\sigma^2} \stackrel{H_0}{\sim} \chi^2(n).$$

3. UE for σ^2 and MS in ANOVA table

(1) MS in ANOVA table I

Define $MSE = \frac{SSE}{n-r}$, $MSM = \frac{SSM}{r-1}$, $MSTO = \frac{C.SSTO}{n-1}$. Then MSE is an UE for σ^2 , MSM and MSTO are UEs for σ^2 under H_0 in Model M_1 .

Proof. $E\left(\frac{SSE}{\sigma^2}\right) = E[\chi^2(n-r)] = n-r \implies E(MSE) = \sigma^2$.

$$E\left(\frac{SSM}{\sigma^2}\right) \stackrel{H_0}{=} E[\chi^2(r-1)] = r-1 \implies E(MSM) \stackrel{H_0}{=} \sigma^2.$$

$$E\left(\frac{C.SSTO}{\sigma^2}\right) = E[\chi^2(n-1)] = n-1 \implies E(MSTO) = \sigma^2.$$

(2) MS in ANOVA table II

Define $MSE = \frac{SSE}{n-r}$, $MSM = \frac{SSM}{r}$, $MSTO = \frac{U.SSTO}{n}$. Then MSE is an UE for σ^2 , MSM and MSTO are UEs for σ^2 under $H_0 : \beta = 0$ in Model M_0 .

Proof. Skipped.

(3) Updated ANOVA tables

Source	SS	DF	MS
Model	SSM	$r - 1$	MSM
Error	SSE	$n - r$	MSE
C.Total	C.SSTO	$n - 1$	MSTO

Source	SS	DF	MS
Model	SSM	r	MSM
Error	SSE	$n - r$	MSE
U.Total	U.SSTO	n	MSTO

L19 F -distributions in ANOVA tables

1. F -distribution in ANOVA table I

(1) F -distributions

Recall: $X^2 \sim \chi^2(r_1)$ and $Y^2 \sim \chi^2(r_2)$ are independent $\implies F = \frac{X^2/r_1}{Y^2/r_2} \sim F(r_1, r_2)$.

$X \sim N(\mu, \Sigma)$, $A' = A$, $B' = B$ and $A\Sigma B = 0 \implies X'AX$ and $X'BX$ are independent.

(2) ANOVA table I

For Model M: $Y \sim N(X\beta, \sigma^2\Sigma)$ with $1_n \in \mathcal{R}(X)$ and $\text{rank}(X) = r$ there exists H_0 .

Under H_0 M is reduced to $M_1 : Y \sim N(1_n\gamma, \sigma^2\Sigma)$. So $\Sigma^{-1/2}Y \stackrel{H_0}{\sim} N(\Sigma^{-1/2}1_n\gamma, \sigma^2I_n)$.

For this model M and H_0 ,

Source	SS	DF	MS	F	$Pr > F$
Model	SSM	$r - 1$	MSM	MSM/MSE	$P(F(r - 1, n - r) > F_{ob})$
Error	SSE	$n - r$	MSE		
C.Total	C.SSTO	$n - 1$			

(3) F -distribution in Table I

$$F = \frac{\text{MSM}}{\text{MSE}} \stackrel{H_0}{\sim} F(r - 1, n - r).$$

Proof. $F = \frac{\text{MSM}}{\text{MSE}} = \frac{\frac{\text{SSM}}{\sigma^2}/(r-1)}{\frac{\text{SSE}}{\sigma^2}/(n-r)}$ where $\frac{\text{SSM}}{\sigma^2} \stackrel{H_0}{\sim} \chi^2(r - 1)$ and $\frac{\text{SSE}}{\sigma^2} \sim \chi^2(n - r)$.

To have $F \stackrel{H_0}{\sim} F(r - 1, n - r)$ we need the independence of $\frac{\text{SSM}}{\sigma^2}$ and $\frac{\text{SSE}}{\sigma^2}$ under H_0 .

But $\frac{\text{SSM}}{\sigma^2} = (\Sigma^{-1/2}Y)'A(\Sigma^{-1/2}Y)$ where $A = \frac{(\Sigma^{-1/2}X)(\Sigma^{-1/2}X)^+ + (\Sigma^{-1/2}1_n)(\Sigma^{-1/2}1_n)^+}{\sigma^2}$

and $\frac{\text{SSE}}{\sigma^2} = (\Sigma^{-1/2}Y)'B(\Sigma^{-1/2}Y)$ where $B = \frac{I - (\Sigma^{-1/2}X)(\Sigma^{-1/2}X)^+}{\sigma^2}$.

Clearly, $A\sigma^2I_nB = 0$. So $\frac{\text{SSM}}{\sigma^2}$ and $\frac{\text{SSE}}{\sigma^2}$ are independent under H_0 .

Conclusion follows.

Comment: We now have completed ANOVA table in (2)

2. F -distribution in ANOVA table II

(1) ANOVA table II

Model M: $Y \sim N(X\beta, \sigma^2\Sigma)$ with $1_n \notin \mathcal{R}(X)$ and $\text{rank}(X) = r$ under $H_0 : \beta = 0$ is

reduced to $M_0 : Y \sim N(0, \sigma^2\Sigma)$. So $\Sigma^{-1/2}Y \stackrel{H_0}{\sim} N(0, \sigma^2I_n)$. For this Model M and

$H_0 : \beta = 0$,

Source	SS	DF	MS	F	$Pr > F$
Model	SSM	r	MSM	MSM/MSE	$P(F(r, n - r) > F_{ob})$
Error	SSE	$n - r$	MSE		
U.Total	U.SSTO	n			

(2) F -distribution in ANOVA table II

$$F = \frac{\text{MSM}}{\text{MSE}} \stackrel{H_0}{\sim} F(r, n - r).$$

Proof. $F = \frac{\text{MSM}}{\text{MSE}} = \frac{\frac{\text{SSM}}{\sigma^2}/(r)}{\frac{\text{SSE}}{\sigma^2}/(n-r)}$ where $\frac{\text{SSM}}{\sigma^2} \stackrel{H_0}{\sim} \chi^2(r)$ and $\frac{\text{SSE}}{\sigma^2} \sim \chi^2(n - r)$.

To have $F \stackrel{H_0}{\sim} F(r, n - r)$ we need the independence of $\frac{\text{SSM}}{\sigma^2}$ and $\frac{\text{SSE}}{\sigma^2}$ under H_0 .

But $\frac{\underline{\text{SSM}}}{\sigma^2} = (\Sigma^{-1/2}Y)'A(\Sigma^{-1/2}Y)$ where $A = \frac{(\Sigma^{-1/2}X)(\Sigma^{-1/2}X)^+}{\sigma^2}$ and $\frac{\underline{\text{SSE}}}{\sigma^2} = (\Sigma^{-1/2}Y)'B(\Sigma^{-1/2}Y)$ where $B = \frac{I - (\Sigma^{-1/2}X)(\Sigma^{-1/2}X)^+}{\sigma^2}$.
Clearly, $A\sigma^2I_nB = 0$. So $\frac{\underline{\text{SSM}}}{\sigma^2}$ and $\frac{\underline{\text{SSE}}}{\sigma^2}$ are independent under H_0 .
Conclusion follows.

Comment: (1) gives a completed ANOVA table.

3. Computation for ANOVA tables

(1) Regression with intercept

Regression with intercept $y = \beta_0 + \beta_1x_1 + \beta_2x_2 + \epsilon$ has data $Y \sim N(X\beta, \sigma^2I_n)$ where $X \in R^{n \times 3}$ has rank 3, and its first column is 1_n . This model is reduced to $Y \sim N(1_n\beta_0, \sigma^2I_n)$ under $H_0 : \beta_1 = 0$ and $\beta_2 = 0$. Thus the model with H_0 has ANOVA table I.

```
data a;
  infile "D:\ex.txt";
  input y x1 x2 @@;
proc reg;
  model y=x1 x2;
run;
```

SAS output displays the ANOVA Table I.

(2) Regression without intercept

Regression without intercept $y = \beta_1x_1 + \beta_2x_2 + \epsilon$ has data $Y \sim N(X\beta, \sigma^2I_n)$. Here $X \in R^{n \times 2}$ has full column rank, but $1_n \notin \mathcal{R}(X)$. Under $H_0 : \beta_1 = 0$ and $\beta_2 = 0$ the reduced model is $Y \sim N(0, \sigma^2I_n)$. Thus the model with H_0 has ANOVA Table II

```
proc reg;
  model y=x1 x2/noint;
run;
```

SAS output displays the ANOVA Table II.

(3) One-way ANOVA

$y = \mu_1I_1 + \mu_2I_2 + \mu_3I_3 + \epsilon$ is ANOVA model with one factor and three levels. Here I_i is the indicator for the i th level, i.e., $I_i = \begin{cases} 0 & \text{if } y \text{ is not the response to level } i \text{ of the factor} \\ 1 & \text{if } y \text{ is the response to level } i \text{ of the factor} \end{cases}$.

This model has data $Y \sim N(J\mu, \sigma^2I_n)$. Here $J = \begin{pmatrix} 1_{n_1} & 0 & 0 \\ 0 & 1_{n_2} & 0 \\ 0 & 0 & 1_{n_3} \end{pmatrix}$, $\mu = \begin{pmatrix} \mu_1 \\ \mu_2 \\ \mu_3 \end{pmatrix}$.

Under $H_0 : \mu_1 = \mu_2 = \mu_3$ the model is reduced to $Y \sim N(1_n\mu_1, \sigma^2I_n)$. So it has ANOVA table I.

```
data a;
  infile "D:\ex3.txt";
  input y Sname $ @@;
proc anova;
  class Sname;
  model y=Sname;
run;
```

SAS output displays the ANOVA Table I.