

POWER OF THE F-TEST

- Suppose C is a $q \times p$ matrix such that $C\beta$ is testable.
- We have established that

$$F = \frac{(C\hat{\beta} - d)' [C(X'X)^{-1}C']^{-1} (C\hat{\beta} - d)/q}{\hat{\sigma}^2}$$

$$\sim F_{q,n-r}(\delta^2) \text{ where}$$

$$\delta^2 = \frac{(C\beta - d)' [C(X'X)^{-1}C']^{-1} (C\beta - d)}{\sigma^2}.$$

- Let $F_{q,n-r,1-\alpha}$ denote the $1 - \alpha$ quantile of the central F distribution with q and $n - r$ d.f.
- The power of the significance level α test of

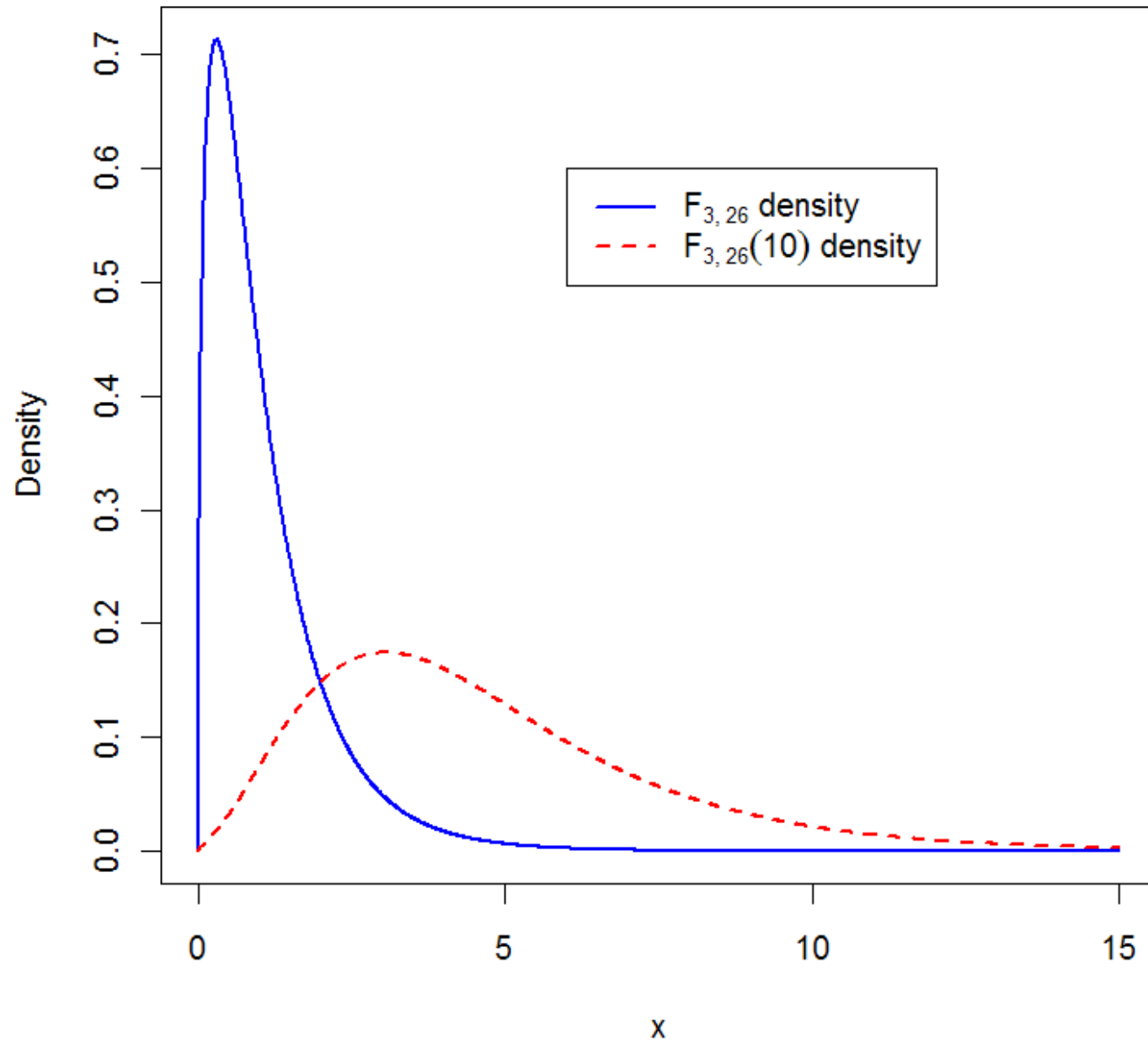
$$H_0 : \mathbf{C}\boldsymbol{\beta} = \mathbf{d}$$

is given by

$$P(F \geq F_{q,n-r,1-\alpha}).$$

- The power is an increasing function of the noncentrality parameter δ^2 .

A Comparison of Central and Non-Central F Densities



```
x=seq(0,15,by=.01)
y1=df(x,3,26)
y2=df(x,3,26,ncp=10)

plot(c(x,x),c(y1,y2),pch=" ",
     xlab="x",ylab="Density",
     main="A Comparison of Central and Non-Central
          F Densities")

lines(x,y1,col="blue",lwd=2)
lines(x,y2,lty=2,col="red",lwd=2)

legend(6,.6,
      c(expression(paste(F[list(3,26)]," density")),
        expression(paste(F[list(3,26)](10)," density"))),
      lty=1:2, col=c("blue","red"), lwd=2)
```

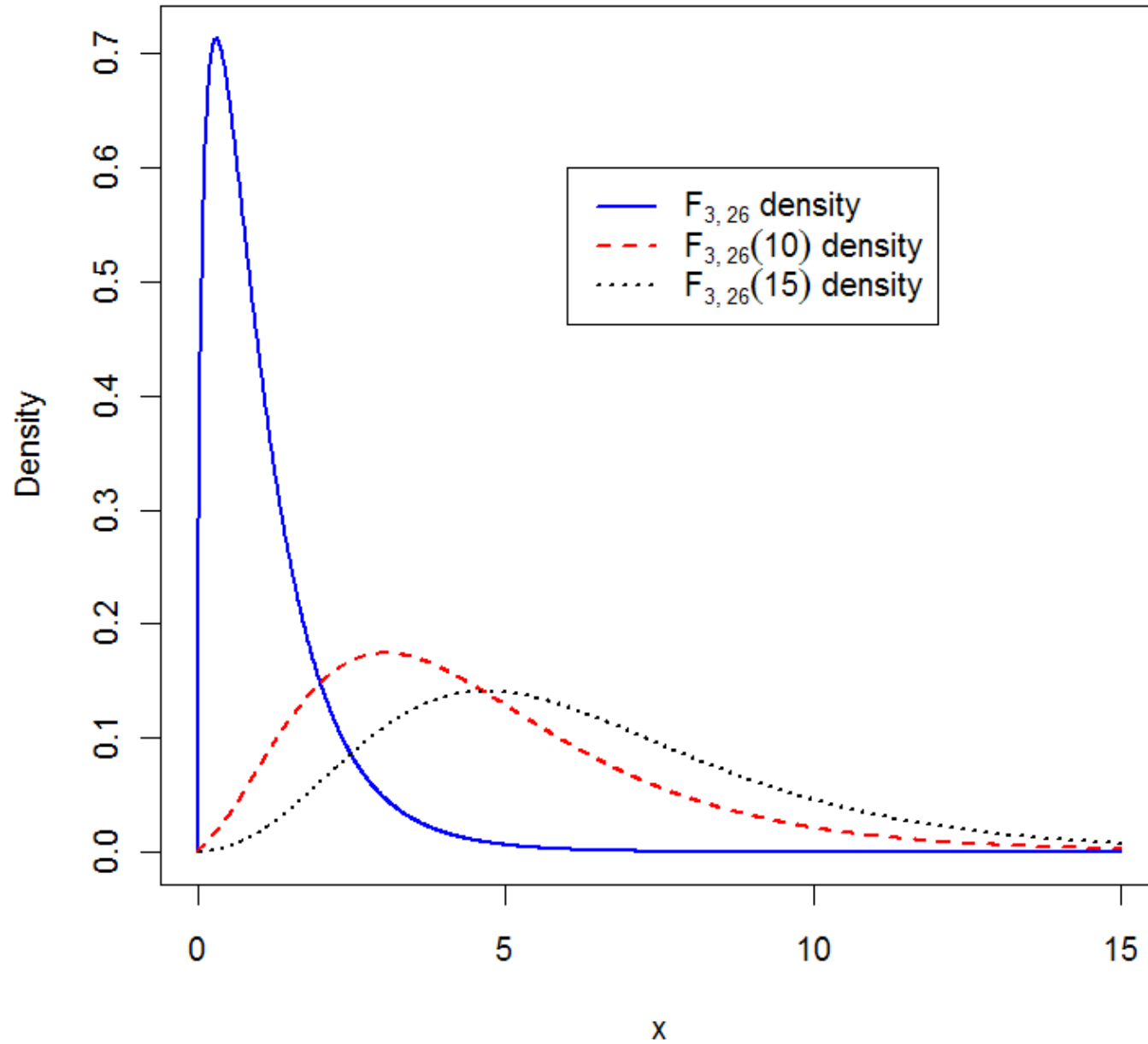
`qf(.95, 3, 26)`

`2.975154`

`1-pf(qf(.95, 3, 26), 3, 26, 10)`

`0.6895487`

A Comparison of Central and Non-Central F Densities



```
plot(c(x,x),c(y1,y2),pch=" ",
     xlab="x",ylab="Density",
     main="A Comparison of Central and Non-Central F
          Densities")

lines(x,y1,col="blue",lwd=2)
lines(x,y2,lty=2, col="red",lwd=2)
lines(x,df(x,3,36,15),lty=3, col="black", lwd=2)

legend(6,.6,
       c(expression(paste(F[list(3,26)]," density")),
         expression(paste(F[list(3,26)](10)," density")),
         expression(paste(F[list(3,26)](15)," density"))),
       lty=1:3, col=c("blue","red","black"), lwd=2)

1-pf(qf(.95,3,26),3,26,15)
```

0.8675232

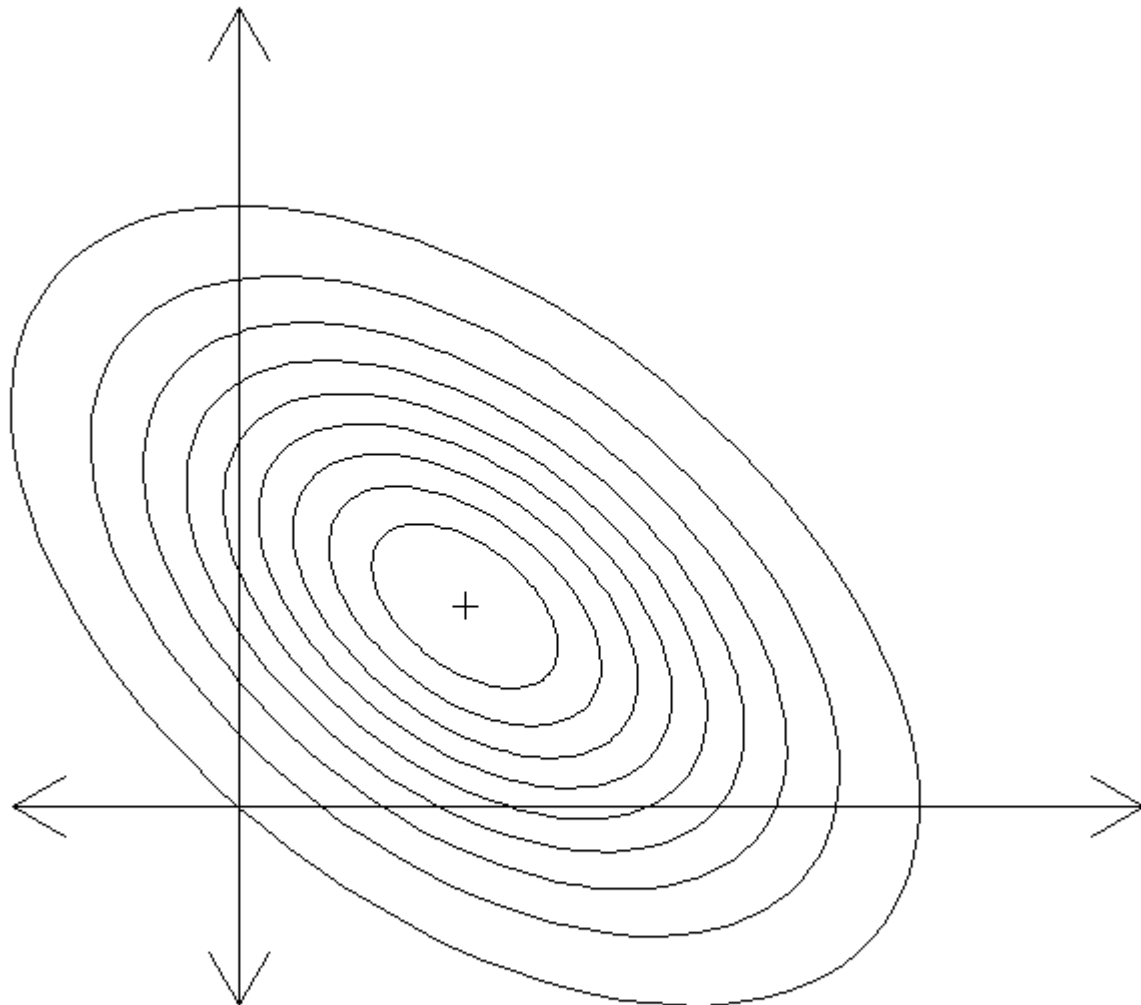
Interpretation of the Noncentrality Parameter

The noncentrality parameter

$$\delta^2 = \frac{(\mathbf{C}\boldsymbol{\beta} - \mathbf{d})' [\mathbf{C}(\mathbf{X}'\mathbf{X})^{-1}\mathbf{C}']^{-1} (\mathbf{C}\boldsymbol{\beta} - \mathbf{d})}{\sigma^2}$$

quantifies the discrepancy between $\mathbf{C}\boldsymbol{\beta}$ and \mathbf{d} with respect to

$$\text{Var}(\mathbf{C}\hat{\boldsymbol{\beta}}) = \sigma^2\mathbf{C}(\mathbf{X}'\mathbf{X})^{-1}\mathbf{C}'.$$



All else being equal, the noncentrality parameter

$$\delta^2 = \frac{(\mathbf{C}\boldsymbol{\beta} - \mathbf{d})' [\mathbf{C}(\mathbf{X}'\mathbf{X})^{-1}\mathbf{C}']^{-1} (\mathbf{C}\boldsymbol{\beta} - \mathbf{d})}{\sigma^2}$$

increases as . . .

- the distance between $\mathbf{C}\boldsymbol{\beta}$ and \mathbf{d} increases,
- σ^2 decreases,
- the design, as defined by \mathbf{X} , improves (e.g., sample size increases).

- Consider a completely randomized design (CRD) with three treatments.
- Suppose $n = 30$ experimental units are available.
- Let n_i denote the number of experimental units assigned to treatment i for $i = 1, 2, 3$.
- Suppose we consider the model

$$Y_{ij} = \mu_i + \epsilon_{ij} \quad i = 1, 2, 3; \quad j = 1, \dots, n_i; \quad \epsilon \sim N(\mathbf{0}, \sigma^2 \mathbf{I})$$

- Consider the test of $H_0 : \mu_1 = \mu_2 = \mu_3$.
- This null hypothesis is equivalent to $H_0 : \mathbf{C}\boldsymbol{\beta} = \mathbf{0}$, where

$$\mathbf{C} = \begin{bmatrix} 1 & -1 & 0 \\ 0 & 1 & -1 \end{bmatrix} \text{ and } \boldsymbol{\beta} = \begin{bmatrix} \mu_1 \\ \mu_2 \\ \mu_3 \end{bmatrix}.$$

- (Actually, \mathbf{C} could be any 2×3 matrix with row space equal to the row space of the \mathbf{C} matrix above.)

$$\mathbf{X}'\mathbf{X} = \begin{bmatrix} n_1 & 0 & 0 \\ 0 & n_2 & 0 \\ 0 & 0 & n_3 \end{bmatrix} \quad (\mathbf{X}'\mathbf{X})^{-1} = \begin{bmatrix} \frac{1}{n_1} & 0 & 0 \\ 0 & \frac{1}{n_2} & 0 \\ 0 & 0 & \frac{1}{n_3} \end{bmatrix}$$

$$\mathbf{C}(\mathbf{X}'\mathbf{X})^{-1}\mathbf{C}' = \begin{bmatrix} \frac{1}{n_1} + \frac{1}{n_2} & -\frac{1}{n_2} \\ -\frac{1}{n_2} & \frac{1}{n_2} + \frac{1}{n_3} \end{bmatrix}$$

$$[\mathbf{C}(\mathbf{X}'\mathbf{X})^{-1}\mathbf{C}']^{-1} = \begin{bmatrix} \frac{1}{n_2} + \frac{1}{n_3} & \frac{1}{n_2} \\ \frac{1}{n_2} & \frac{1}{n_1} + \frac{1}{n_2} \end{bmatrix} \frac{1}{\frac{1}{n_1 n_2} + \frac{1}{n_1 n_3} + \frac{1}{n_2 n_3}}$$

$$= \begin{bmatrix} \frac{1}{n_2} + \frac{1}{n_3} & \frac{1}{n_2} \\ \frac{1}{n_2} & \frac{1}{n_1} + \frac{1}{n_2} \end{bmatrix} \frac{n_1 n_2 n_3}{n}$$

- Suppose $\mu_1 = 3, \mu_2 = 2, \mu_3 = 1$, and $\sigma^2 = 1$

- Then $C\beta - d = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ and

$$\begin{aligned} \delta^2 &= [1, 1] \begin{bmatrix} \frac{1}{n_2} + \frac{1}{n_3} & \frac{1}{n_2} \\ \frac{1}{n_2} & \frac{1}{n_1} + \frac{1}{n_2} \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} \frac{n_1 n_2 n_3}{n} \\ &= \frac{4n_1 n_3 + n_1 n_2 + n_2 n_3}{n}. \end{aligned}$$

• Now suppose $\mu_1 = 3, \mu_2 = \mu_3 = 1, \sigma^2 = 1$

• Then $C\beta - \mathbf{d} = \begin{bmatrix} 2 \\ 0 \end{bmatrix}$ and

$$\begin{aligned} \delta^2 &= [2, 0] \begin{bmatrix} \frac{1}{n_2} + \frac{1}{n_3} & \frac{1}{n_2} \\ \frac{1}{n_2} & \frac{1}{n_1} + \frac{1}{n_2} \end{bmatrix} \begin{bmatrix} 2 \\ 0 \end{bmatrix} \frac{n_1 n_2 n_3}{n} \\ &= \frac{4n_1(n_2 + n_3)}{n}. \end{aligned}$$

μ_1	μ_2	μ_3	n_1	n_2	n_3	δ^2
3	2	1	5	20	5	10.0
3	2	1	10	10	10	20.0
3	2	1	12	6	12	24.0
3	1	1	5	20	5	16. $\bar{6}$
3	1	1	10	10	10	26. $\bar{6}$
3	1	1	12	6	12	28.8