

General Linear Hypothesis Test

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Notation: Write $n \times (p+1)$ matrix $\mathbf{X} = (\mathbf{1}, \mathbf{x}_1, \dots, \mathbf{x}_p)$, where $\mathbf{1}$ is n -dimensional vector with all entries being 1 and $\mathbf{x}_j \in \mathbb{R}^n$. Denote the column space of \mathbf{X} by $\mathcal{C}_{\mathbf{X}}$, i.e. $\mathcal{C}_{\mathbf{X}} = \{c_0\mathbf{1} + c_1\mathbf{x}_1 + \dots + c_p\mathbf{x}_p : c_0, c_1, \dots, c_p \in \mathbb{R}\}$. For $\mathbf{Y} \in \mathbb{R}^n$, let $\Pi(\mathbf{Y}|\mathcal{C}_{\mathbf{X}})$ denote the projection of \mathbf{Y} onto $\mathcal{C}_{\mathbf{X}}$. Note \mathbf{X} is full rank as in usual assumption. Suppose \mathbf{A} is linear subspace in \mathbb{R}^k . Let $\mathbf{A}^\perp = \{\mathbf{x} \in \mathbb{R}^n : \mathbf{x}^t \mathbf{a} = \mathbf{0} \text{ for all } \mathbf{a} \in \mathbf{A}\}$.

Let \mathbf{A} be $(p+1) \times q$ matrix of full column rank and \mathbf{c} be q -dimensional vector ($q < p+1$). Suppose we want to test

$$H_0 : \mathbf{A}^t \boldsymbol{\beta} = \mathbf{c} \text{ versus } H_1 : \mathbf{A}^t \boldsymbol{\beta} \neq \mathbf{c}$$

Under the constraint $\mathbf{A}^t \boldsymbol{\beta} = \mathbf{c}$, we have to find minimizer of $g(\boldsymbol{\beta}) = \|\mathbf{Y} - \mathbf{X}\boldsymbol{\beta}\|^2$. Let $\hat{\boldsymbol{\beta}}_r = \arg \min_{\mathbf{A}^t \boldsymbol{\beta} = \mathbf{c}} \|\mathbf{Y} - \mathbf{X}\boldsymbol{\beta}\|^2$.

We first consider the case when $\mathbf{c} = \mathbf{0}$.

Claim: Let \mathbf{Z} be $q \times n$ matrix. Regard \mathbf{Z} as a linear map $\mathbf{Z} : \mathbb{R}^n \rightarrow \mathbb{R}^q$. Then, $\text{Null}(\mathbf{Z}) = \text{Range}(\mathbf{Z}^t)^\perp$ so that $\mathbb{R}^n = \text{Range}(\mathbf{Z}^t) \oplus \text{Null}(\mathbf{Z})$.

Proof. Take any $\mathbf{v} \in \mathbb{R}^q$. If $\mathbf{u} \in \text{Null}(\mathbf{Z})$, $\mathbf{u}^t \mathbf{Z}^t \mathbf{v} = (\mathbf{Z}\mathbf{u})^t \mathbf{v} = \mathbf{0}^t \mathbf{v} = \mathbf{0}$. Hence $\text{Null}(\mathbf{Z}) \subseteq \text{Range}(\mathbf{Z}^t)^\perp$. Now choose any $\mathbf{w} \in \text{Range}(\mathbf{Z}^t)^\perp$. Then $\mathbf{w}^t (\mathbf{Z}^t \mathbf{v}) = (\mathbf{Z}\mathbf{w})^t \mathbf{v} = \mathbf{0}$ for any $\mathbf{v} \in \mathbb{R}^q$. Since this holds for all $\mathbf{v} \in \mathbb{R}^q$, $\mathbf{Z}\mathbf{w} = \mathbf{0}$ and thus $\mathbf{w} \in \text{Null}(\mathbf{Z})$, which implies $\text{Range}(\mathbf{Z}^t)^\perp \subseteq \text{Null}(\mathbf{Z})$. Therefore, we conclude that $\text{Null}(\mathbf{Z}) = \text{Range}(\mathbf{Z}^t)^\perp$. \square

Assuming $\mathbf{c} = \mathbf{0}$, $\mathbf{A}^t \boldsymbol{\beta} = \mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t \mathbf{X} \boldsymbol{\beta} = \mathbf{0}$. Note that because \mathbf{A} and \mathbf{X} are full rank, $\mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A}$ is invertible. Using the result of claim, one can clearly see that $\mathbf{X}\hat{\boldsymbol{\beta}}_r$ is the projection of \mathbf{Y} onto

$$\mathcal{C}_{\mathbf{X}} \cap \text{null}(\mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t) = \mathcal{C}_{\mathbf{X}} \cap \text{Range}(\mathbf{X} (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A})^\perp = \mathcal{C}_{\mathbf{X}} \cap \mathcal{C}_{\mathbf{X} (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A}}^\perp$$

Since $\mathcal{C}_{\mathbf{X} (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A}} \subseteq \mathcal{C}_{\mathbf{X}}$,

$$\begin{aligned} \mathbf{X}\hat{\boldsymbol{\beta}}_r &= \Pi(\mathbf{Y}|\mathcal{C}_{\mathbf{X}} \cap \mathcal{C}_{\mathbf{X} (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A}}^\perp) \\ &= \Pi(\mathbf{Y}|\mathcal{C}_{\mathbf{X}}) - \Pi(\mathbf{Y}|\mathcal{C}_{\mathbf{X}} \cap \mathcal{C}_{\mathbf{X} (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A}}) \\ &= \Pi(\mathbf{Y}|\mathcal{C}_{\mathbf{X}}) - \Pi(\mathbf{Y}|\mathcal{C}_{\mathbf{X} (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A}}) \\ &= \mathbf{X}(\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t \mathbf{Y} - \mathbf{X}(\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A} (\mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t \mathbf{X} (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A})^{-1} \mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t \mathbf{Y} \\ &= \mathbf{X}(\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t \mathbf{Y} - \mathbf{X}(\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A} (\mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A})^{-1} \mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t \mathbf{Y} \end{aligned}$$

Thus, we obtain $\hat{\boldsymbol{\beta}}_r = (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t \mathbf{Y} - (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A} (\mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A})^{-1} \mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t \mathbf{Y} \dots (*)$. Now, we generalize this result for all \mathbf{c} . Since $\mathbf{A}^t \boldsymbol{\beta} = \mathbf{A}^t \mathbf{A} (\mathbf{A}^t \mathbf{A})^{-1} \mathbf{c}$, $\mathbf{A}^t (\boldsymbol{\beta} - \mathbf{A} (\mathbf{A}^t \mathbf{A})^{-1} \mathbf{c}) = \mathbf{0}$, if we let $\boldsymbol{\gamma} = \boldsymbol{\beta} - \mathbf{A} (\mathbf{A}^t \mathbf{A})^{-1} \mathbf{c}$,

$$\|\mathbf{Y} - \mathbf{X}\boldsymbol{\beta}\|^2 = \|\mathbf{Y} - \mathbf{X}(\boldsymbol{\gamma} + \mathbf{A} (\mathbf{A}^t \mathbf{A})^{-1} \mathbf{c})\|^2 = \|\mathbf{Y} - \mathbf{X}\mathbf{A} (\mathbf{A}^t \mathbf{A})^{-1} \mathbf{c} - \mathbf{X}\boldsymbol{\gamma}\|^2$$

with $\mathbf{A}^t \boldsymbol{\gamma} = \mathbf{0}$. Because $\hat{\boldsymbol{\beta}}_r = \hat{\boldsymbol{\gamma}} + \mathbf{A} (\mathbf{A}^t \mathbf{A})^{-1} \mathbf{c}$, it suffices to find $\hat{\boldsymbol{\gamma}}$, where $\hat{\boldsymbol{\gamma}} = \arg \min_{\mathbf{A}^t \boldsymbol{\gamma} = \mathbf{0}} \|\mathbf{Y} - \mathbf{X}\mathbf{A} (\mathbf{A}^t \mathbf{A})^{-1} \mathbf{c} - \mathbf{X}\boldsymbol{\gamma}\|^2$. But $\hat{\boldsymbol{\gamma}}$ can be easily found by replacing \mathbf{Y} with $\mathbf{Y} - \mathbf{X}\mathbf{A} (\mathbf{A}^t \mathbf{A})^{-1} \mathbf{c}$ in (*). Hence,

$$\begin{aligned}
\hat{\beta}_r &= \hat{\gamma} + \mathbf{A}(\mathbf{A}^t \mathbf{A})^{-1} \mathbf{c} \\
&= (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t (\mathbf{Y} - \mathbf{X} \mathbf{A}(\mathbf{A}^t \mathbf{A})^{-1} \mathbf{c}) - (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A}(\mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A})^{-1} \mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t (\mathbf{Y} - \mathbf{X} \mathbf{A}(\mathbf{A}^t \mathbf{A})^{-1} \mathbf{c}) \\
&\quad + \mathbf{A}(\mathbf{A}^t \mathbf{A})^{-1} \mathbf{c} \\
&= (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t \mathbf{Y} - \mathbf{A}(\mathbf{A}^t \mathbf{A})^{-1} \mathbf{c} - (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A}(\mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A})^{-1} \mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t (\mathbf{Y} - \mathbf{X} \mathbf{A}(\mathbf{A}^t \mathbf{A})^{-1} \mathbf{c}) \\
&\quad + \mathbf{A}(\mathbf{A}^t \mathbf{A})^{-1} \mathbf{c} \\
&= (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t \mathbf{Y} - (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A}(\mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A})^{-1} \mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t (\mathbf{Y} - \mathbf{X} \mathbf{A}(\mathbf{A}^t \mathbf{A})^{-1} \mathbf{c}) \\
&= \hat{\beta} - (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A}(\mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A})^{-1} \mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t (\mathbf{Y} - \mathbf{X} \mathbf{A}(\mathbf{A}^t \mathbf{A})^{-1} \mathbf{c})
\end{aligned}$$

where $\hat{\beta} = (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t \mathbf{Y}$. To derive test, note that if $\mathbf{A}^t \beta = \mathbf{c}$ is not true, $\mathbf{R}(\beta_{-r} | \beta_r) = \|\mathbf{Y} - \mathbf{X} \hat{\beta}_r\|^2 - \|\mathbf{Y} - \mathbf{X} \hat{\beta}\|^2$ tends to get larger. Let $\hat{\mathbf{u}} = (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A}(\mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A})^{-1} \mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t (\mathbf{Y} - \mathbf{X} \mathbf{A}(\mathbf{A}^t \mathbf{A})^{-1} \mathbf{c})$.

$$\begin{aligned}
\mathbf{R}(\beta_{-r} | \beta_r) &= \|\mathbf{Y} - \mathbf{X} \hat{\beta} + \mathbf{X} \hat{\mathbf{u}}\|^2 - \|\mathbf{Y} - \mathbf{X} \hat{\beta}\|^2 \\
&= \|\mathbf{Y} - \mathbf{X} \hat{\beta}\|^2 + 2(\mathbf{X} \hat{\mathbf{u}})^t (\mathbf{Y} - \mathbf{X} \hat{\beta}) + \|\mathbf{X} \hat{\mathbf{u}}\|^2 - \|\mathbf{Y} - \mathbf{X} \hat{\beta}\|^2 \\
&= 2\hat{\mathbf{u}}^t \mathbf{X}^t (\mathbf{Y} - \mathbf{X} \hat{\beta}) + \|\mathbf{X} \hat{\mathbf{u}}\|^2 = 2\hat{\mathbf{u}}^t \mathbf{X}^t (\mathbf{I} - \mathbf{X}(\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t) \mathbf{Y} + \|\mathbf{X} \hat{\mathbf{u}}\|^2 \\
&= 2\hat{\mathbf{u}}^t (\mathbf{X}^t - \mathbf{X}^t) \mathbf{Y} + \|\mathbf{X} \hat{\mathbf{u}}\|^2 = \|\mathbf{X} \hat{\mathbf{u}}\|^2
\end{aligned}$$

With simple calculation,

$$\begin{aligned}
\|\mathbf{X} \hat{\mathbf{u}}\|^2 &= (\mathbf{Y} - \mathbf{X} \mathbf{A}(\mathbf{A}^t \mathbf{A})^{-1} \mathbf{c})^t \mathbf{X}(\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A}(\mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A})^{-1} \mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t (\mathbf{Y} - \mathbf{X} \mathbf{A}(\mathbf{A}^t \mathbf{A})^{-1} \mathbf{c}) \\
&= (\mathbf{A}^t \hat{\beta} - \mathbf{c})^t (\mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A})^{-1} (\mathbf{A}^t \hat{\beta} - \mathbf{c}) = \|(\mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A})^{-\frac{1}{2}} (\mathbf{A}^t \hat{\beta} - \mathbf{c})\|^2
\end{aligned}$$

We know that $\hat{\beta} \sim N_{p+1}(\beta, \sigma^2 (\mathbf{X}^t \mathbf{X})^{-1})$. Hence, $\mathbf{A}^t \hat{\beta} \sim N_q(\mathbf{A}^t \beta, \sigma^2 \mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A})$. Under H_0 , $\mathbf{A}^t \hat{\beta} \sim N_q(\mathbf{c}, \sigma^2 \mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A})$. So $(\mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A})^{-\frac{1}{2}} (\mathbf{A}^t \hat{\beta} - \mathbf{c}) / \sigma \sim N_q(\mathbf{0}_q, \mathbf{I}_q)$ and thus $\mathbf{R}(\beta_{-r} | \beta_r) / \sigma^2 \sim \chi^2(q)$.

In the lecture, we've seen that $\text{SSE} / \sigma^2 = \|\mathbf{Y} - \mathbf{X} \hat{\beta}\|^2 / \sigma^2 \sim \chi^2(n - p - 1)$. By direct computation, it can be verified that

$$\begin{aligned}
\|\mathbf{Y} - \mathbf{X} \hat{\beta}\|^2 &= \mathbf{Y}^t (\mathbf{I} - \mathbf{X}(\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t) \mathbf{Y} \\
&= (\mathbf{Y} - \mathbf{X} \mathbf{A}(\mathbf{A}^t \mathbf{A})^{-1} \mathbf{c})^t (\mathbf{I} - \mathbf{X}(\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t) (\mathbf{Y} - \mathbf{X} \mathbf{A}(\mathbf{A}^t \mathbf{A})^{-1} \mathbf{c})
\end{aligned}$$

Recall that if $\mathbf{z} \sim N_k(\mu, \Sigma)$ and \mathbf{P}, \mathbf{Q} are $k \times k$ symmetric, idempotent matrices, $\mathbf{z}^t \mathbf{P} \mathbf{z}$ and $\mathbf{z}^t \mathbf{Q} \mathbf{z}$ are independent if and only if $\mathbf{P} \Sigma \mathbf{Q} = \mathbf{0}$. Because

$$\begin{aligned}
&\mathbf{X}(\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A}(\mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A})^{-1} \mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t (\mathbf{I} - \mathbf{X}(\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t) \\
&= \mathbf{X}(\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A}(\mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A})^{-1} \mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} (\mathbf{X}^t - \mathbf{X}^t) = \mathbf{0}
\end{aligned}$$

and $\mathbf{X}(\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A}(\mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{A})^{-1} \mathbf{A}^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t$, $\mathbf{I} - \mathbf{X}(\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t$ are symmetric, idempotent, $\mathbf{R}(\beta_{-r} | \beta_r) / \sigma^2$ and SSE / σ^2 are independent. Therefore,

$$F \equiv \frac{\frac{\mathbf{R}(\beta_{-r} | \beta_r)}{q \sigma^2}}{\frac{\text{SSE}}{(n-p-1) \sigma^2}} = \frac{\mathbf{R}(\beta_{-r} | \beta_r) / q}{\text{SSE} / (n - p - 1)} \sim F(q, n - p - 1)$$

So we reject H_0 if $F > F_\alpha(q, n - p - 1)$.