

Rank of products of matrices

A and B matrices, AB defined. Then
 $\text{rank}(AB) \leq \min(\text{rank}(A), \text{rank}(B))$.

Rank and eigenvalues

$A_{n \times n}$ symmetric implies $\text{rank}(A)$ equals the number of nonzero eigenvalues.

$A_{n \times n}$ symmetric and idempotent with rank r implies r eigenvalues are 1 and $n-r$ are zero.

$A_{n \times n}$ symmetric and idempotent implies $\text{tr}(A) = \text{rank}(A)$.

Independence of quadratic forms

A and B symmetric and idempotent with $AB=0$ and

$X \sim N_n(\mu, \sigma^2 I)$. Then $(X - \mu)^T A (X - \mu)$ and $(X - \mu)^T B (X - \mu)$ are independent.

Distribution of quadratic forms

$A_{n \times n}$ symmetric and idempotent with rank r . $X \sim N_n(\mu, \sigma^2 I)$.

Then $\frac{(X - \mu)^T A (X - \mu)}{\sigma^2} \sim \chi^2(r)$.

Multiple linear regression model.

$Y = X\beta + \varepsilon$, where $E[\varepsilon] = 0$, and $\text{Cov}(\varepsilon) = \sigma^2 I$

Least squares estimators

Coefficients: $\hat{\beta} = (X^T X)^{-1} X^T Y$

Model: $\hat{Y} = X \hat{\beta}$

Residuals: $Y - \hat{Y}$

Properties

$$E[\hat{\beta}] = \beta, \quad Cov[\hat{\beta}] = \sigma^2 (X^T X)^{-1}$$