

Stat243: Homework 7

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1 Problem 1

We compare the standard error $\frac{s}{\sqrt{n}}$ with effect size to see if it correctly characterizes the uncertainty of the estimated regression coefficient.

2 Problem 2

$$\begin{aligned}
 A &= \Gamma \Lambda \Gamma^T \\
 Az &= \Gamma \Lambda (\Gamma^T z) \\
 Az^T(Az) &= z^T A^T A z = z^T \Gamma \Lambda \Gamma^T \Gamma \Lambda \Gamma^T z = z^T \Gamma \Lambda \Lambda \Gamma^T z \\
 y &= \Gamma^T z \\
 &= y^T \Lambda^2 y \\
 \|y\|_2^2 &= \sum (\Gamma_{i1} z_1 + \dots + \Gamma_{in} z_n)^2 = \sum \Gamma_{i1}^2 z_1^2 + \dots + \Gamma_{in}^2 z_n^2 + 2 \sum_{k \neq l} \Gamma_{ik} \Gamma_{il} z_i z_l \\
 \|y\|_2^2 &= z_1^2 + \dots + z_n^2 = 1 \\
 y^T \Gamma^2 y &= \sum y_i^2 \Gamma_i^2 \\
 \|A\|_2 &= \sup \sqrt{\sum y_i^2 \Lambda_i^2} \\
 \max \Gamma_{i..n}^2 &= \Gamma_k^2 \\
 \|A\|_2 &= \|\Gamma_k\|
 \end{aligned}$$

Therefore, the norm A is the largest of the absolute value of the eigenvalues of A for symmetric A .

3 Problem 3

3.1 3A

$$\begin{aligned}X &= USV^T \\X^T &= VSU^T \\X^TX &= VSU^TUSV^T = VS^2V^T \\X^TXV &= VS^2 \\V^T(X^TX)V &= (XV)^TXV = \|XV\|_2^2 \geq 0\end{aligned}$$

Therefore, X^TX is a positive definite matrix, so the SVD for X^TX is equal to eigendecomposition. VS^2V^T is the SVD of X^TX , based on the definition of eigendecomposition, columns of V are eigenvectors of X^TX . Also, columns of V are right singular vectors of X . In conclusion, the right singular vectors of X are eigenvectors of X^TX . Since S^2 's diagonal elements are eigenvalues of X^TX , S 's diagonal elements are singular values of X , the eigenvalues of X^TX are squares of the singular values of X .

3.2 3B

$$\begin{aligned}Zv &= \lambda v \\(\sum + cI)v &= \lambda v \\\sum v &= (\lambda - c)v\end{aligned}$$

so $\lambda - c$ is the eigenvalue of \sum . The eigenvalues of Z can be computed by adding constant c to the eigenvalues of \sum

4 Problem 4

4.1 Problem 4b

```
Efficient_Solve <- function(X,Y,A,b){
  C = crossprod(X)
  d = crossprod(X,Y)
  tmp1 = crossprod(solve(c,t(A)),t(A))
  tmp2 = crossprod(solve(c, t(A)), d)
  beta_hat = solve(c,d) + solve(c,crossprod(A,solve(tmp1, -tmp2+b)))
  beta_hat
}
```

5 Problem 5

5.1 5A

Instrument variables have an important property: The matrix of correlation between variables in X and the variables in Z is of maximum possible rank, meaning that we want X to retain only the correlation between X and Z . However, since X and Z are sparse matrixes, resulting X is also sparse, which would fail the expectation of stage 1 which is supposed to generate a matrix containing all the correlation items(fitted values) for the second stage.

5.2 5B

Stage 1: Regress each variable in X on Z to obtain a matrix with fitted values.

Stage 2: Regress y on X

Key: Use P_z as an intermediate

$$\begin{aligned} Z(Z^T Z)^{-1} Z^T X &= P_z X \\ \beta &= (\hat{X}^T \hat{X})^{-1} \hat{X}^T y = (X^T P_z^T P_z X)^{-1} X^T P_z^T y = (X^T P_z X)^{-1} X^T P_z y \\ P_z^T P_z &= [Z(Z^T Z)^{-1} Z^T]^T = P_z \end{aligned}$$

Pseudocode:

Compute P_z using spam package

plug P_z into β equation

return β