MATH 437 HW2

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and \sigma_4 = 1. 1.c. A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} A is different from the identity and is orthogonal. 2.a.
 \hat{\beta} = (A^T \Sigma^{-1} A)^{-1} A^T \Sigma^{-1} y \text{ 2.b. } \text{mean} = \hat{\hat{\beta}} = (A^T \Sigma^{-1} A)^{-1} A^T \Sigma^{-1} A \beta E(\hat{\beta}) = \beta Cov(\hat{\beta}) = (A^T \Sigma^{-1} A)^{-1} A^T \Sigma^{-1} A \beta E(\hat{\beta}) = \beta Cov(\hat{\beta}) = (A^T \Sigma^{-1} A)^{-1} A^T \Sigma^{-1} A \beta E(\hat{\beta}) = \beta Cov(\hat{\beta}) = (A^T \Sigma^{-1} A)^{-1} A^T \Sigma^{-1} A \beta E(\hat{\beta}) = \beta Cov(\hat{\beta}) = (A^T \Sigma^{-1} A)^{-1} A^T \Sigma^{-1} A \beta E(\hat{\beta}) = \beta Cov(\hat{\beta}) = (A^T \Sigma^{-1} A)^{-1} A^T \Sigma^{-1} A \beta E(\hat{\beta}) = \beta Cov(\hat{\beta}) = (A^T \Sigma^{-1} A)^{-1} A^T \Sigma^{-1} A \beta E(\hat{\beta}) = \beta Cov(\hat{\beta}) = (A^T \Sigma^{-1} A)^{-1} A \beta E(\hat{\beta}) = \beta Cov(\hat{\beta}) = (A^T \Sigma^{-1} A)^{-1} A \beta E(\hat{\beta}) = (A^T \Sigma^{-1} A)^{-
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                       \begin{bmatrix} a_{1,1} & a_{1,2} \end{bmatrix}
 2.c. Cov(r) = I - A(A^T \Sigma^{-1} A)^{-1} A^T \Sigma^{-1} \Sigma (I - A(A^T \Sigma^{-1} A)^{-1} A^T \Sigma^{-1}) 3.a.
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                       3.b.
  \begin{bmatrix} a_{1,1} & a_{1,1} & a_{1,1} & a_{1,2} & a_{1,2} & a_{1,2} \\ a_{2,1} & a_{2,1} & a_{2,1} & a_{2,2} & a_{2,2} \end{bmatrix} 3.c. \begin{bmatrix} a_{1,1} & a_{1,2} & 0 & 0 \\ a_{2,1} & a_{2,2} & 0 & 0 \\ 0 & 0 & a_{1,1} & a_{1,2} \\ 0 & 0 & a_{2,1} & a_{2,2} \end{bmatrix} 3.d. \begin{bmatrix} a_{1,1} & 0 & a_{1,2} & 0 \\ 0 & a_{1,1} & 0 & a_{1,2} & 0 \\ 0 & 0 & a_{2,1} & a_{2,2} \end{bmatrix}
\begin{bmatrix} a_{1,1}a_{1,1} & a_{1,1}a_{1,2} & a_{1,2}a_{1,1} & a_{1,2}a_{1,2} \\ a_{1,1}a_{2,1} & a_{1,1}a_{2,2} & a_{1,2}a_{2,1} & a_{1,2}a_{2,2} \\ a_{1,1}a_{2,1} & a_{2,1}a_{1,2} & a_{2,2}a_{1,1} & a_{1,2}a_{2,2} \\ a_{2,1}a_{2,1} & a_{2,1}a_{2,2} & a_{2,2}a_{2,1} & a_{2,2}a_{2,2} \end{bmatrix} 3.f. \text{ let } A = \begin{bmatrix} 1 & 2 \\ 3 & 5 \end{bmatrix} \text{ then } (A \otimes A)(I \otimes A) = \begin{bmatrix} 7 & 18 & 21 & 54 \\ 12 & 31 & 36 & 93 \\ 14 & 36 & 35 & 90 \\ 24 & 62 & 60 & 155 \end{bmatrix}
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                     4.a.
 \$E(Y) = (0,0,0,0) \hat{\ } \{T\} \$ \text{ 4.b. } Var(Y) = (2,2,1,1)^T \text{ 4.c. } E(U) = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} Var(U) = \begin{bmatrix} 2 & 2 \\ 1 & 1 \end{bmatrix} \text{ 5.a. } x1 = N(1,1)
  x2=N(1,5) 5.b.
  sigma \leftarrow matrix(c(1,-2,-2,5),nrow=2)
  mu=c(1,1)
  x = matrix(c(1,0,1,-1),nrow=2)
  x%*%mu
  ##
                                                             [,1]
  ## [1,]
  ## [2,]
 t(x)%*%sigma%*%x
                                                            [,1] [,2]
  ## [1,]
                                                            1
  ## [2,]
  5.c.
  x = matrix(c(1,1,1,-1),nrow=2)
  x%*%mu
  ##
                                                             [,1]
```

[1,] ## [2,]

t(x)%*%sigma%*%x

5.d. new
$$\mu = \begin{bmatrix} a_{1,1} & a_{1,2} \\ a_{2,1} & a_{2,2} \end{bmatrix} \mu$$
 new $\sigma = (\begin{bmatrix} a_{1,1} & a_{1,2} \\ a_{2,1} & a_{2,2} \end{bmatrix})^T \Sigma \begin{bmatrix} a_{1,1} & a_{1,2} \\ a_{2,1} & a_{2,2} \end{bmatrix}$ 5.e. new $\mu = \begin{bmatrix} a_{1,1} & a_{1,2} \\ a_{2,1} & a_{2,2} \end{bmatrix} b \mu$ new $\sigma = (\begin{bmatrix} a_{1,1} & a_{1,2} \\ a_{2,1} & a_{2,2} \end{bmatrix} b)^T \Sigma \begin{bmatrix} a_{1,1} & a_{1,2} \\ a_{2,1} & a_{2,2} \end{bmatrix} b$ 5.f. $a = 1, b = -1$ 6.a. Since the vector is positive and the sigma matrix is always semi definite the matrix $a^T \Sigma a$ is always positive definite.