

Problem 1

$$\text{since } \text{Cov}(X) = E((X - E(X))(X - E(X))^T) \quad | \quad \Rightarrow \\ Y = AX$$

$$\begin{aligned} \Rightarrow \text{Cov}(AX) &= E(AX - E(AX))(AX - E(AX))^T \\ &= E(A(X - E(X))(X - E(X))^T A^T) \\ &= AE[(X - E(X))(X - E(X))^T]A^T \\ &\quad \text{``}\Sigma_X\text{''} \\ &= A\Sigma_XA^T \end{aligned}$$

Problem 2

$$\begin{aligned} \text{i) } \text{Cov}(Z, X_1) &= E(ZX_1^T) - E(Z)\underbrace{E(X_1)^T}_{\text{covariance definition}} \quad \rightarrow X_1 \text{ has mean 0} \Rightarrow E(X_1)^T \text{ vector} \\ &= E(ZX_1^T) = E((X_2 - \alpha X_1)X_1^T) \quad \text{row of 0} \\ &= E(X_2 X_1^T) - \alpha E(X_1 X_1^T) = 0 \quad \text{``}\text{Cov}(Z, X_1)\text{''} \\ &\Rightarrow \alpha = \frac{E(X_2 X_1^T)}{E(X_1 X_1^T)} = \frac{\sigma_{21}}{\sigma_{11}} \end{aligned}$$

b) Suppose  $v = \begin{pmatrix} z \\ x_1 \end{pmatrix}$

$$\Sigma = \begin{pmatrix} \Sigma_{zz} & \Sigma_{zx_1} \\ \Sigma_{x_1 z} & \Sigma_{x_1 x_1} \end{pmatrix} \quad \text{Cov}(z, x_1) = 0$$

$$N = \begin{pmatrix} N_z \\ N_{x_1} \end{pmatrix} \quad \text{Cov}(z, x_1) = 0$$

$$f_v(v) = \frac{1}{(2\pi)^{1/2} |\Sigma|^{1/2}} \exp \left\{ -\frac{1}{2} (v - N)^T \Sigma^{-1} (v - N) \right\}$$

$$\Rightarrow (v_1^T - N_1^T \ v_2^T - N_2^T) \begin{pmatrix} \Sigma_{zz} & 0 \\ 0 & \Sigma_{x_1 x_1} \end{pmatrix}^{-1} \begin{pmatrix} v_1 - N_1 \\ v_2 - N_2 \end{pmatrix} =$$

cov matrix diagonal

$$= (v_1^T - N_1^T \ v_2^T - N_2^T) \begin{pmatrix} \Sigma_{11}^{-1} & 0 \\ 0 & \Sigma_{22}^{-1} \end{pmatrix} \begin{pmatrix} v_1 - N_1 \\ v_2 - N_2 \end{pmatrix} =$$

$$= (v_1 - N_1)^T \Sigma_{11}^{-1} (v_1 - N_1) + (v_2 - N_2)^T \Sigma_{22}^{-1} (v_2 - N_2)$$

$\downarrow$   
This can be factored because it is within exponential function

So, the joint pdf  $f_v(v) = f_z(v_1) h_{x_1}(v_2)$  shows that  $z$  and  $x_1$  are independent

$$\begin{aligned} V_{xz}(z) &= \text{Var}(x_2 - \alpha x_1) = \text{Var}(x_2) + \alpha^2 \text{Var}(x_1) - 2\alpha \text{Cov}(x_2, x_1) \\ &= \sigma_{22} + (\sigma_{21} \sigma_{11}^{-1})^2 \sigma_{11} - 2\sigma_{21} \sigma_{11}^{-1} \sigma_{21} \\ &= \sigma_{22} + \sigma_{21} \sigma_{11}^{-1} \sigma_{21} - 2\sigma_{21} \sigma_{11}^{-1} \sigma_{21} \\ &= \sigma_{22} - \sigma_{21} \sigma_{11}^{-1} \sigma_{21} \end{aligned}$$

c)  $E(x_2 | X_1) = \alpha x + E(z) = E(x_2 | X_1) = \alpha x + E(z) - \alpha E(x_1) = \sigma_{21} \sigma_{11}^{-1} x$

$$V_{xz}(x_2 | X_1) = \text{Var}(\alpha x + z | X_1) + \text{Var}(z | X_1) = \text{Var}(z) = \sigma_{22} - \sigma_{21} \sigma_{11}^{-1} \sigma_{21}$$

Problem 3

$$s^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$$

$P = I_n - \frac{1}{n} \mathbf{1}_n \mathbf{1}_n^T$  → symmetric and idempotent matrix (HW1)

a)  $(n-1)s^2 = X^T P X$

$$s^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2 = \frac{1}{n-1} (P X)^T P X$$

$$= \frac{1}{n-1} X^T P^T P X$$

$$= \frac{1}{n-1} X^T P X$$

P symmetric

$$= \frac{1}{n-1} X^T P X$$

P idempotent

$$\Rightarrow s^2(n-1) = X^T P X$$

$$\begin{aligned} b) E(\sum (x_i - \bar{x})^2) &= \sum (x_i - \bar{x})^2 + n(\bar{x} - \mu)^2 \\ &= \sum ((x_i - \bar{x}) + (\bar{x} - \mu))^2 \\ &= \sum ((x_i - \bar{x})^2 + 2(x_i - \bar{x})(\bar{x} - \mu) + (\bar{x} - \mu)^2) \\ &= \sum (x_i - \bar{x})^2 + 2(\bar{x} - \mu) \sum (x_i - \bar{x}) + n(\bar{x} - \mu)^2 \end{aligned}$$

$$= \sum (x_i - \bar{x})^2 + n(\bar{x} - \mu)^2$$

$$\begin{aligned} E(s^2) &= E\left(\frac{1}{n-1} \sum (x_i - \bar{x})^2\right) = \frac{1}{n-1} E\left(\sum ((x_i - \bar{x})^2 - n(\bar{x} - \mu)^2)\right) \\ &= \frac{1}{n-1} \left( \sum E((x_i - \bar{x})^2) - n E((\bar{x} - \mu)^2) \right) = \frac{1}{n-1} \left( \sum \sigma^2 - n \text{Var}(\bar{x}) \right) \end{aligned}$$

$$= \frac{1}{n-1} \left( (n-1)\sigma^2 - \frac{n\sigma^2}{n} \right) = \frac{1}{n-1} (n-1)\sigma^2 = \sigma^2$$

#Ceibal10

c)  $\text{Var}(s^2)$        $\text{Var}_{\text{normal}} \rightarrow \sigma^2$

$$\text{Var}\left(\frac{(n-1)s^2}{\sigma^2}\right) = \frac{(n-1)}{\sigma^4} \text{Var}(s^2) = 2(n-1)$$

$$\text{Var}(s^2) = \frac{2\sigma^4}{n-1}$$

d)  $\sum_{i=1}^n (x_i - N)^2 = \sum_{i=1}^n (x_i - \bar{x} + \bar{x} - N)^2 = \sum_{i=1}^n (x_i - \bar{x})^2 + 2(x_i - \bar{x})(\bar{x} - p) + (\bar{x} - N)^2$

(1) (2) (3)

$$= \sum_{i=1}^n (x_i - \bar{x})^2 + 2(\bar{x} - N) \sum_{i=1}^n (x_i - \bar{x}) + n(\bar{x} - N)^2$$

$$= \sum_{i=1}^n (x_i - \bar{x})^2 + n(\bar{x} - N)^2$$

(1)  $\rightarrow \frac{\sum_{i=1}^n (x_i - N)^2}{\sigma^2}$  has a chi-squared distribution w/  $n$  degrees of freedom  
because it is the sum of  $n$  independent standard normals

(2)  $\rightarrow \frac{n(\bar{x} - p)^2}{\sigma^2}$  has chi-squared distribution w/ 1 degree of freedom  
because it is the square of a standard normal.

e) Since  $s^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$  is a function of  $x_i - \bar{x}, i=1, \dots, n$   
it follows that  $s^2$  is independent of  $\bar{x}$

## Problem 5

a)

Coefficients:					
	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	22.55565	17.19680	1.312	0.1968	
sex	-22.11833	8.21111	-2.694	0.0101 *	
status	0.05223	0.28111	0.186	0.8535	
income	4.96198	1.02539	4.839	1.79e-05 ***	
verbal	-2.95949	2.17215	-1.362	0.1803	
---					

Sex and income are statistically significant at the 5% level.

> confint(lm, level=0.95)	
	2.5 % 97.5 %
(Intercept)	-12.1489038 57.2602050
sex	-38.6890301 -5.5476301
status	-0.5150722 0.6195399
income	2.8926538 7.0313047
verbal	-7.3430703 1.4240833

- b) Sex is just a dummy variable representing the categorical data of sex. 1 represents female and 0 represents male.

c)

Analysis of Variance Table					
Model 1: gamble ~ income					
Model 2: gamble ~ sex + status + income + verbal					
Res.Df RSS Df Sum of Sq F Pr(>F)					
1 45 28009					
2 42 21624 3 6384.8 4.1338 0.01177 *					
---					
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1					

Since the p-value is very small the null hypothesis is rejected.

d)

Analysis of Variance Table					
Model 1: gamble ~ 1					
Model 2: gamble ~ sex + status + income + verbal					
Res.Df RSS Df Sum of Sq F Pr(>F)					
1 46 45689					
2 42 21624 4 24066 11.686 1.815e-06 ***					
---					
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1					

Problem 6

$$t \text{ value} = \frac{\text{estimate}}{\text{std. error}} \Rightarrow t \text{ value} = \frac{0.7190}{0.1535} = 4.684$$

$$p\text{-value} \rightarrow 2 \cdot p(t > \text{abs}(t), df=28) = 6.5776 \times 10^{-5}$$

$$R^2 = 1 - \frac{RSS}{TSS}$$

$$F = \frac{\frac{TSS - RSS}{r-1}}{\frac{RSS}{n-r}} = 21.94, r=2$$

$$RSE = \sqrt{\frac{RSS}{n-r}} \Rightarrow 0.8662 = \sqrt{\frac{RSS}{28}} \Rightarrow RSS = 21.008$$

$$F = \frac{TSS - 21}{\frac{21}{28}} = 21.94 \Rightarrow TSS = 37.455$$

$$\Rightarrow R^2 = 1 - \frac{21}{37.455} = 0.5606$$

## Appendix

```
# title: "MSDS596 - HW2"
# author: "Diego Sarachaga"
# date: "10/02/2018"

library(faraway)
data(teengamb)

lm <- lm(gamble ~ sex + status + income + verbal, teengamb)
summary(lm)

#a
# Coefficients:
#          Estimate Std. Error t value Pr(>|t|)
# (Intercept) 22.55565 17.19680 1.312 0.1968
# sex        -22.11833  8.21111 -2.694 0.0101 *
# status       0.05223  0.28111  0.186 0.8535
# income      4.96198  1.02539  4.839 1.79e-05 ***
# verbal     -2.95949  2.17215 -1.362 0.1803
# ---
# Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 '' 1
#
# Residual standard error: 22.69 on 42 degrees of freedom
# Multiple R-squared: 0.5267,    Adjusted R-squared: 0.4816
# F-statistic: 11.69 on 4 and 42 DF, p-value: 1.815e-06

#Sex and income are statistically significant at the 5% level
confint(lm, level=0.95)

#Confident intervals
#sex      -38.6890301 -5.5476301
#income    2.8926538  7.0313047

#b
#Sex is a dummy variable representing the categorical data of sex. A 1
represents female and a 0 represents male.

#c
lmi <- lm(gamble ~ income, teengamb)
summary(lmi)

anova(lmi, lm)
# Analysis of Variance Table
```

```
#  
# Model 1: gamble ~ income  
# Model 2: gamble ~ sex + status + income + verbal  
# Res.Df RSS Df Sum of Sq F Pr(>F)  
# 1 45 28009  
# 2 42 21624 3 6384.8 4.1338 0.01177 *  
# ---  
# Signif. codes: 0 '****' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 '' 1
```

#Since the p-value is so small the null hypothesis is rejected.

```
#d  
nullmod <- lm(gamble ~ 1, teengamb)  
anova(nullmod, lm)  
# Analysis of Variance Table  
#  
# Model 1: gamble ~ 1  
# Model 2: gamble ~ sex + status + income + verbal  
# Res.Df RSS Df Sum of Sq F Pr(>F)  
# 1 46 45689  
# 2 42 21624 4 24066 11.686 1.815e-06 ***  
# ---  
# Signif. codes: 0 '****' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 '' 1
```