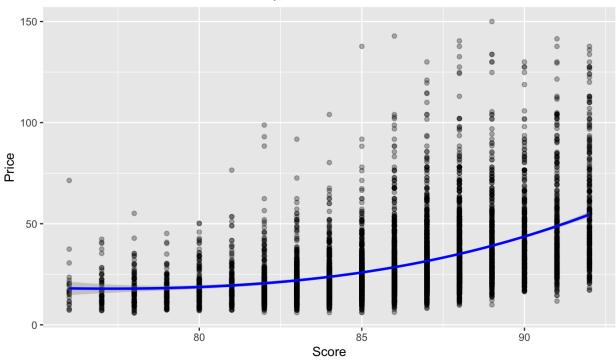
ECOM20001 Econometrics 1 Assignment 2

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Q1

The relationship between price and score appears to be non-linear as the fitted line is by inspection not straight (linear). Visually, there are no inflection points within the domain (the range of the attribute 'Score' in the wine data).





Q2

Price = -197.188 + 2.653 Score

Price = 1,330.419 - 33.230 Score + 0.210 Score^2

Price = -1,704.065 + 74.251 Score - 1.057 Score^2 + 0.005 Score^3

Detailed version:

		Dependent variable:	
	(1)	Price (2)	(3)
score	2.653*** (0.064)	-33.230*** (2.369)	74.251 (67.408)
score2		0.210*** (0.014)	-1.057 (0.802)
score3			0.005 (0.003)
Constant	-197.188*** (5.388)	1,330.419*** (99.432)	-1,704.065 (1,886.575)
Observations R2 Adjusted R2 Residual Std. Error F Statistic	6,979 0.221 0.221 16.609 (df = 6977) 1,981.860*** (df = 1; 6977)		
Note:		*	fp<0.1; **p<0.05; ***p<0.01

sequential hypothesis

- Model 3:

H0: coef(score3) = 0, H1: coef(score3) != 0

The p value of coefficient of score3 >0.05, hence failing to reject H0 at 5% significance level, which means the cubic relationship between price and score is not statistically significant when tested against a quadratic fitted relationship. Therefore, we process to the next level of testing — quadratic model.

Model 2:

H0: coef(score2) = 0, H1: coef(score2) != 0

The p value of coefficient of score2 <0.05, hence we reject H0 at 5% significance level, which means the quadratic relationship between price and score is statistically significant when tested against a linearly fitted relationship. Hence, we stop the sequential hypothesis testing, and conclude that a quadratic relationship is appropriate for modeling the relationship between price and score.

	Col 1 coef	Col 1 SE	Col 2 coef	Col 2 SE	Col 3 coef	Col 3 SE
Constant	1330.419	99.432	1250.858	98.386	1363.257	89.152
Score	-33.230	2.369	-31.170	2.343	-33.431	2.125
Score squared	0.210	0.014	0.198	0.014	0.208	0.013
Wine type pinotnoir			-3.903	0.934	-3.381	0.897
Wine type cabernet			-3.365	0.961	-3.969	0.866
Wine type merlot			-8.069	0.918	-8.282	0.828
Wine type syrah			-11.429	0.992	-11.898	0.939

Note: for column 3, the wine region and year dummies are controlled but not included in the table above.

		Dependent variable:	
	(1)	Price (2)	(3)
score	-33.230*** (2.369)	-31.170*** (2.343)	-33.431*** (2.125)
score2	0.210*** (0.014)	0.198*** (0.014)	0.208*** (0.013)
pinotnoir		-3.903*** (0.934)	-3.381*** (0.897)
cabernet		-3.365*** (0.961)	-3.969*** (0.866)
merlot		-8.069*** (0.918)	-8.282*** (0.828)
syrah		-11.428*** (0.992)	-11.898*** (0.939)

Constant	1,330.419*** (99.432)	1,250.858*** (98.386)	1,363.257*** (89.152)
Observations	6,979	6,979	6,979
R2	0.249	0.275	0.415
Adjusted R2	0.249	0.274	0.413
Residual Std. Error	16.309 (df = 6976)	16.031 (df = 6972)	14.415 (df = 6955)
F Statistic	1,157.354*** (df = 2; 6976)	440.614*** (df = 6; 6972)	214.724*** (df = 23; 6955)
 Note:		:	*p<0.1; **p<0.05; ***p<0.01

Q4

H0: coefficient (score²) = 0 H1: coefficient (score²) \neq 0

Reject the hypothesis that the coefficient on score^2 is 0 at 5% significance level, because the p-value 2.2e-16 < 0.05. Hence we conclude that the non-linear relationship between price and score is statistically significant in column 3.

Test statistic = 271.55 following F distribution on F 1, 6955

Q5

Test regression coefficient on [wine types] pinotnoir, cabernet, merlot and syrah are jointly equal to each other

H0: coefficient(pinotnoir) = coefficient(cabernet) = coefficient(merlot) = coefficient(syrah)

H1: at least one of the above condition does not hold

P-value = 2.2e-16 < 0.05, hence we reject the hypothesis that coefficients on [wine types] pinotnoir, cabernet, merlot and syrah are jointly equal to each other. We conclude that wine types have statistically significant different impacts on a wine's price at 5% significance level, holding other factors fixed.

Test statistic = 115.91 following F distribution on 3, 6955 df

Q6

	score changes from price=80 to price=85	score changes from price=85 to price=90
Partial effect on price (\$)	6.1	16.6
SE	0.389	0.551
Confidence interval	[5.350, 6.850]	[15.520, 17.680]

The partial effect on the price of changing score from 80 to 85 is \$6.1 which is lower than that from 85 to 90 that is \$16.6. Even though the price change in both cases is \$5, there is a greater partial effect because of the nonlinearity nature of the relationship as demonstrated in previous questions. The sequential testing shows it is statistically significant to say that the relationship between price and score is quadratic. As foreshadowed/shown in the Question 1 Graph, as score increases, the partial effect on price of changing one unit of score also increases; which corresponds to the results from this question.

Q7

	constant	log(score)		Wine type cabernet	Wine type merlot	Wine type syrah
Coefficie nt	-22.256	5.709	-0.050	-0.107	-0.193	-0.294
SE	0.608	0.137	0.022	0.020	0.020	0.024

elasticity of price with respect to score = 5.71%

	Dependent variable:
	log(Price)
log(score)	5.709***
	(0.137)
pinotnoir	-0.050**
	(0.022)
cabernet	-0.107***
	(0.020)
merlot	-0.193***
	(0.020)
syrah	-0.294***
	(0.024)
Constant	-22.256***
	(0.608)
Observations	6,979
R2	0.487
Adjusted R2	0.486
Residual Std. Error F Statistic	0.378 (df = 6956) 300.642*** (df = 22; 6956)
	Jew. 042*** (dr = 22; 6956)
Note:	*p<0.1; **p<0.05; ***p<0.01

Note: for this table, the wine region and year dummies are controlled but not included in the table above.

Q8

H0: coefficient(log(score)) = 1; H1: coefficient(log(score)) != 1

P-value = 2.2e-16 < 0.05, hence we reject the hypothesis that coefficient of log(score) = 1. We conclude that the elasticity of price with respect to score is not equal to 1% at 5% significance level.

Test statistic = 1179.5 following F distribution on 1, 6956 df.

Q9

	Col 1 coef	Col 1 SE	Col 2 coef	Col 2 SE
Constant	-22.256	0.608	-27.395	2.427
log(score)	5.709	0.137	6.859	0.543
Wine type pinotnoir	-0.050	0.022	5.872	2.668
Wine type cabernet	-0.107	0.020	0.338	2.580
Wine type merlot	-0.193	0.020	11.452	2.575
Wine type syrah	-0.294	0.024	11.367	3.087
Interaction between log(score) and wine type pinotnoir			-1.326	0.597
Interaction between log(score) and wine type cabernet			-0.098	0.577
Interaction between log(score) and wine type			-2.613	0.576

merlot			
Interaction between log(score) and wine type syrah		-2.610	0.691

	Dependen ⁻	t variable:
		rice
	(1)	(2)
log(score)	5.709***	6.859***
	(0.137)	(0.543)
pinotnoir	-0.050**	5.872**
	(0.022)	(2.668)
cabernet	-0.107***	0.338
	(0.020)	(2.580)
merlot	-0.193***	11.452***
	(0.020)	(2.575)
syrah	-0.294***	11.367***
	(0.024)	(3.087)
og(score):pinotnoir		-1.326**
		(0.597)
og(score):cabernet		-0.098
		(0.577)
og(score):merlot		-2.613***
		(0.576)
og(score):syrah		-2.610***
		(0.691)
Constant	-22.256***	-27.395***
	(0.608)	(2.427)
 Observations	 6,979	6,979
32	0.487	0.494
Adjusted R2	0.486	0.492 0.376 (df = 6952)
Statistic	0.378 (df = 6956) 300.642*** (df = 22; 6956)	0.376 (df = 6952) $260.631*** (df = 26; 6952)$
 lote:		*p<0.1; **p<0.05; ***p<0.01

The elastic w respect to wine type pinotnoir = 6.859-1.326 = 5.533 %The elastic w respect to wine type cabernet = 6.859-0.098 = 6.761 % The elast w respect to wine type merlot = 6.859-2.613 = 4.246 %The elast w respect to wine type syrah = 6.859-2.601 = 4.258 %

Q10

H0: coef(log(score)*pinotnoir) = coef(log(score)*cabernet) = coef(log(score)*merlot) = coef(log(score)*syrah)

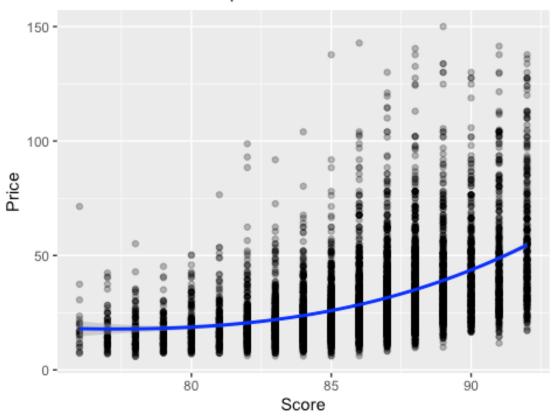
H1: one or more of above are not equal to each other

P-value = 2.2e-16 < 0.05, hence we reject the hypothesis that coefficient of interaction of log(score) and pinotnoir = coefficient of interaction of log(score) and cabernet = coefficient of interaction of log(score) and merlot = coefficient of interaction of log(score) and syrah. We conclude that the elasticity of price with respect to score is not the same for the 4 wine types at 5% significance level.

Test statistic = 26.781 following F distribution on 3, 6952 df.

```
library(stargazer)
## Please cite as:
## Hlavac, Marek (2022). stargazer: Well-Formatted Regression and Summary Statistics Tables.
## R package version 5.2.3. https://CRAN.R-project.org/package=stargazer
library(AER)
## Loading required package: car
## Loading required package: carData
## Loading required package: lmtest
## Loading required package: zoo
## Attaching package: 'zoo'
## The following objects are masked from 'package:base':
##
       as.Date, as.Date.numeric
## Loading required package: sandwich
## Loading required package: survival
library(ggplot2)
## Warning in register(): Can't find generic `scale_type` in package ggplot2 to
## register S3 method.
data = read.csv('as3_wine.csv')
Q1.
ggplot(data, aes(y=price, x=score)) +
 geom_point(alpha=.3) +
 stat_smooth(method = 'lm', formula= y ~poly(x,3), col='blue') +
 ggtitle('Relationship Between Price and Score') +
 theme(plot.title = element_text(hjust = 0.5)) +
  scale_x_continuous(name='Score') +
 scale_y_continuous(name='Price')
```

Relationship Between Price and Score



Q2.

```
# create nonlinear variables
data$score2 = data$score * data$score
data$score3 = data$score2 * data$score
# fit linear regressions
lm2_1 = lm(price ~ score, data = data)
lm2_2 = lm(price ~ score + score2, data = data)
lm2_3 = lm(price ~ score + score2 + score3, data = data)
# get covar and se
cov2_1 = vcovHC(lm2_1, type="HC1")
cov2_2 = vcovHC(lm2_2, type="HC1")
cov2_3 = vcovHC(lm2_3, type="HC1")
se2_1 = sqrt(diag(cov2_1))
se2_2 = sqrt(diag(cov2_2))
se2_3 = sqrt(diag(cov2_3))
# get stargazer summary
stargazer(lm2_1, lm2_2, lm2_3, type='text',
          se = list(se2_1, se2_2, se2_3),
          digits = 3,
          dep.var.labels = c("Price"),
            out='Q2_output.txt')
```

```
______
                                              Dependent variable:
##
                                                     Price
                         (1)
                                                                             (3)
## score
                          2.653***
                                                  -33.230***
                                                                             74.251
##
                            (0.064)
                                                    (2.369)
(67.408)
##
## score2
                                                    0.210***
                                                                             -1.057
##
                                                     (0.014)
(0.802)
##
## score3
                                                                              0.005
##
(0.003)
##
## Constant
                         -197.188***
                                               1,330.419***
1,704.065
                           (5.388)
                                                   (99.432)
(1,886.575)
                                                     6,979
## Observations
                            6,979
                                                                              6,979
## R2
                             0.221
                                                     0.249
                                                                              0.249
                                                     0.249
## Adjusted R2
                             0.221
                                                                              0.249
## Residual Std. Error 16.609 (df = 6977) 16.309 (df = 6976) 16.308 (df =
## F Statistic 1,981.860*** (df = 1; 6977) 1,157.354*** (df = 2; 6976) 772.648*** (df
= 3; 6975)
## Note:
                                                                    *p<0.1; **p<0.05;
***p<0.01
Q3.
# get the nonlinear regressions
lm3_1 = lm(price ~ score + score2, data = data)
lm3_2 = lm(price ~ score + score2 + pinotnoir + cabernet + merlot + syrah, data = data)
lm3_3 = lm(price ~ score + score2 + pinotnoir + cabernet + merlot + syrah + napa + bayarea +
sonoma + scoast + carn + sierra + mendo + wash + d1991 + d1992 + d1993 + d1994 + d1995 + d1996
+ d1997 + d1998 + d1999, data = data)
# get covar and se
cov3_1 = vcovHC(lm3_1, type="HC1")
cov3_2 = vcovHC(lm3_2, type="HC1")
cov3_3 = vcovHC(lm3_3, type="HC1")
se3_1 = sqrt(diag(cov3_1))
se3_2 = sqrt(diag(cov3_2))
se3_3 = sqrt(diag(cov3_3))
```

##

```
# get stargazer summary
stargazer(lm3_1, lm3_2, lm3_3, type='text',
         se = list(se3_1, se3_2, se3_3),
         digits = 3,
         dep.var.labels = c("Price"),
           out='Q3_output.txt')
##
##
_______
##
                                                  Dependent variable:
##
                                                        Price
                                 (1)
                                                                                   (3)
                             -33.230***
                                                       -31.170***
## score
33.431***
##
                               (2.369)
                                                         (2.343)
                                                                                 (2.125)
##
                              0.210***
                                                       0.198***
                                                                                 0.208***
## score2
##
                               (0.014)
                                                        (0.014)
                                                                                 (0.013)
##
                                                        -3.903***
                                                                                -3.381***
## pinotnoir
##
                                                        (0.934)
                                                                                 (0.897)
##
                                                                                -3.969***
                                                        -3.365***
## cabernet
##
                                                        (0.961)
                                                                                 (0.866)
##
                                                                                -8.282***
                                                        -8.069***
## merlot
##
                                                         (0.918)
                                                                                 (0.828)
##
                                                       -11.428***
## syrah
11.898***
                                                         (0.992)
                                                                                 (0.939)
##
##
## napa
                                                                                14.335***
##
                                                                                 (0.637)
##
                                                                                 9.192***
## bayarea
##
                                                                                 (0.856)
##
                                                                                 7.329***
## sonoma
                                                                                 (0.584)
##
##
                                                                                 5.060***
## scoast
##
                                                                                 (0.606)
##
## carn
                                                                                 6.483***
##
                                                                                 (0.752)
##
                                                                                 4.839***
## sierra
##
                                                                                 (1.114)
##
                                                                                 2.819***
## mendo
                                                                                 (0.810)
##
##
                                                                                  -0.002
## wash
                                                                                 (0.648)
##
```

```
##
## d1991
10.661***
##
                                                                                      (0.834)
##
## d1992
10.244***
##
                                                                                      (0.858)
##
## d1993
                                                                                     -9.089***
##
                                                                                     (0.823)
##
                                                                                     -8.892***
## d1994
##
                                                                                     (0.789)
##
## d1995
                                                                                     -5.273***
##
                                                                                     (0.801)
##
## d1996
                                                                                     -3.600***
##
                                                                                     (0.811)
##
## d1997
                                                                                      -1.164
##
                                                                                      (0.814)
##
## d1998
                                                                                     4.530***
##
                                                                                      (0.906)
##
## d1999
                                                                                     1.844**
                                                                                      (0.845)
## Constant
                            1,330.419***
                                                1,250.858***
1,363.257***
                                (99.432)
                                                           (98.386)
                                                                                     (89.152)
                                                                                      6,979
## Observations
                                 6,979
                                                            6,979
                                 0.249
                                                             0.275
                                                                                      0.415
## Adjusted R2
                                 0.249
                                                             0.274
                                                                                      0.413
## Residual Std. Error 16.309 (df = 6976)
                                                   16.031 (df = 6972)
                                                                                14.415 (df =
## F Statistic
                      1,157.354*** (df = 2; 6976) 440.614*** (df = 6; 6972) 214.724*** (df =
23; 6955)
## Note:
                                                                            *p<0.1; **p<0.05;
***p<0.01
```

Q4)

```
# run linear hypothesis to test for non linearity
linearHypothesis(lm3_3, c('score2=0'), vcov = vcovHC(lm3_3, 'HC1'))

## Linear hypothesis test
##
## Hypothesis:
## score2 = 0
##
## Model 1: restricted model
## Model 2: price ~ score + score2 + pinotnoir + cabernet + merlot + syrah +
```

```
##
       napa + bayarea + sonoma + scoast + carn + sierra + mendo +
##
       wash + d1991 + d1992 + d1993 + d1994 + d1995 + d1996 + d1997 +
##
       d1998 + d1999
##
## Note: Coefficient covariance matrix supplied.
##
##
    Res.Df Df
                   F
                        Pr(>F)
## 1
       6956
      6955 1 271.55 < 2.2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
Q5)
# Test the specificed regression coefficients are jointly equal to each other
linearHypothesis(lm3 3, c('pinotnoir = cabernet', 'cabernet = merlot', 'merlot = syrah'), vcov
= vcovHC(lm3 3, 'HC1'))
## Linear hypothesis test
##
## Hypothesis:
## pinotnoir - cabernet = 0
## cabernet - merlot = 0
## merlot - syrah = 0
##
## Model 1: restricted model
## Model 2: price ~ score + score2 + pinotnoir + cabernet + merlot + syrah +
##
       napa + bayarea + sonoma + scoast + carn + sierra + mendo +
##
       wash + d1991 + d1992 + d1993 + d1994 + d1995 + d1996 + d1997 +
       d1998 + d1999
##
##
## Note: Coefficient covariance matrix supplied.
##
     Res.Df Df
##
                    F
                        Pr(>F)
## 1
       6958
## 2
      6955 3 115.91 < 2.2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
Q6)
# partial effect on price when score changes from 80 to 85
print('partial effects')
## [1] "partial effects"
(dprice6_1 = (-33.43 * (85) + 0.21 * (85**2)) - (-33.43 * (80) + 0.21 * (80**2)))
## [1] 6.1
# partial effect on price when score changes from 85 to 90
(dprice6_2 = (-33.43 * (90) + 0.21 * (90**2)) - (-33.43 * (85) + 0.21 * (85**2)))
## [1] 16.6
# get the changes in quadratic term
85**2 - 80**2
## [1] 825
90**2 - 85**2
```

```
## [1] 875
# get f statistics for the joint test of H0 = 5*score + 825* score^2 when score changes from
80 to 85
Ftest6 1 = linearHypothesis(lm3 3, c('5*score + 825* score2'), vcov = vcovHC(lm3 3, "HC1"))
Fstat6 1 = Ftest6_1[2,3]
# get f statistics for the joint test of H0 = 5*score + 875* score^2 when score changes from
85 to 90
Ftest6 2 = linearHypothesis(lm3 3, c('5*score + 875* score2'), vcov = vcovHC(lm3 3, "HC1"))
Fstat6 2 = Ftest6 2[2,3]
# get SE when score changes from 80 to 85
print('SE')
## [1] "SE"
(se_dprice6_1 = abs(dprice6_1)/sqrt(Fstat6_1))
## [1] 0.3828554
# get SE when score changes from 85 to 90
(se_dprice6_2 = abs(dprice6_2)/sqrt(Fstat6_2))
## [1] 0.5509774
# get 95% CI for the partial effects
print('CI')
## [1] "CI"
(dprice_ci95_6_1 = dprice6_1 + c(-1.96, 1.96) * se_dprice6_1)
## [1] 5.349603 6.850397
(dprice_{ci95_6_2} = dprice6_2 + c(-1.96, 1.96) * se_dprice6_2)
## [1] 15.52008 17.67992
Q7)
# Run LM with log of these two variables
lm7_1 = lm(log(price) ~ log(score) + pinotnoir + cabernet + merlot + syrah + napa + bayarea +
sonoma + scoast + carn + sierra + mendo + wash + d1991 + d1992 + d1993 + d1994 + d1995 + d1996
+ d1997 + d1998 + d1999, data = data)
# get covar and se
cov7_1 = vcovHC(lm7_1, type="HC1")
se7_1 = sqrt(diag(cov7_1))
# get stargazer summary
stargazer(lm7_1, type='text',
         se = list(se7_1),
         digits = 3,
         dep.var.labels = c("log(Price)"),
           out='Q7_output.txt')
##
##
                          Dependent variable:
```

##		
##		log(Price)
	log(score)	5.709***
##		(0.137)
##		-0.050**
##	pinotnoir	(0.022)
##		(0.022)
##	cabernet	-0.107***
##		(0.020)
##	merlot	-0.193***
##	iller Tot	(0.020)
##		(
	syrah	-0.294***
## ##		(0.024)
	napa	0.582***
##		(0.021)
##		
	bayarea	0.422***
##		(0.027)
	sonoma	0.386***
##	,	(0.021)
##		
	scoast	0.332***
## ##		(0.022)
	carn	0.355***
##		(0.026)
##		0.252***
##	sierra	0.262*** (0.040)
##		(0.040)
	mendo	0.208***
##		(0.030)
##	wash	0.160***
##	waSII	(0.024)
##		(3.024)
	d1991	-0.334***
##		(0.024)
##	d1992	-0.343***
##	W + 2 2 2	(0.024)
##		
	d1993	-0.296***
##		(0.023)
##	d1994	-0.247***
##	G I J J T	(0.022)
##		
	d1995	-0.144***
##		(0.022)
##	d1996	-0.103***
##	G1000	(0.022)
##		
##	d1997	-0.048**

##

```
##
                               (0.022)
##
                              0.089***
## d1998
##
                               (0.023)
##
                               0.038*
## d1999
##
                               (0.022)
##
                             -22.256***
## Constant
##
                               (0.608)
##
## Observations
## R2
                                0.487
## Adjusted R2
                               0.486
## Residual Std. Error 0.378 (df = 6956)
## F Statistic 300.642*** (df = 22; 6956)
## Note: *p<0.1; **p<0.05; ***p<0.01
Q8)
# run linear hypothesis to test equivalency of effects
linearHypothesis(lm7_1, c('log(score) = 1'), vcov = vcovHC(lm7_1, 'HC1'))
## Linear hypothesis test
##
## Hypothesis:
## log(score) = 1
##
## Model 1: restricted model
## Model 2: log(price) ~ log(score) + pinotnoir + cabernet + merlot + syrah +
##
      napa + bayarea + sonoma + scoast + carn + sierra + mendo +
##
      wash + d1991 + d1992 + d1993 + d1994 + d1995 + d1996 + d1997 +
##
      d1998 + d1999
##
## Note: Coefficient covariance matrix supplied.
##
##
   Res.Df Df
                 F Pr(>F)
## 1 6957
## 2 6956 1 1179.5 < 2.2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Q9)
# Run LM with log of these two variables and interaction
lm9_1 = lm(log(price) ~ log(score) + log(score)*pinotnoir + log(score)*cabernet +
log(score)*merlot + log(score)*syrah + pinotnoir + cabernet + merlot + syrah + napa + bayarea
+ sonoma + scoast + carn + sierra + mendo + wash + d1991 + d1992 + d1993 + d1994 + d1995 +
d1996 + d1997 + d1998 + d1999, data = data)
# get covar and se
cov9_1 = vcovHC(lm9_1, type="HC1")
se9 1 = sqrt(diag(cov9 1))
# get stargazer summary
stargazer(lm7_1, lm9_1, type='text',
         se = list(se7_1, se9_1),
         digits = 3,
```

```
dep.var.labels = c("Price"),
           out='Q9_output.txt')
## -----
                                        Dependent variable:
##
##
                                               Price
                                 (1)
                                                            (2)
## log(score)
                                5.709***
                                                          6.859***
##
                                (0.137)
                                                          (0.543)
##
## pinotnoir
                                -0.050**
                                                           5.872**
##
                                (0.022)
                                                           (2.668)
                               -0.107***
## cabernet
                                                           0.338
##
                                (0.020)
                                                          (2.580)
##
                               -0.193***
## merlot
                                                         11.452***
##
                                (0.020)
                                                          (2.575)
##
                               -0.294***
                                                         11.367***
## syrah
##
                                (0.024)
                                                          (3.087)
##
                                0.582***
                                                          0.582***
## napa
                                (0.021)
                                                          (0.021)
                                0.422***
                                                          0.422***
## bayarea
##
                                (0.027)
                                                          (0.027)
##
                                0.386***
                                                          0.388***
## sonoma
##
                                (0.021)
                                                          (0.021)
##
                                0.332***
                                                          0.333***
## scoast
##
                                (0.022)
                                                          (0.022)
##
                                0.355***
                                                          0.356***
## carn
##
                                (0.026)
                                                          (0.026)
##
                                0.262***
                                                          0.267***
## sierra
##
                                (0.040)
                                                          (0.040)
##
                                0.208***
                                                          0.212***
## mendo
##
                                (0.030)
                                                           (0.030)
##
                                0.160***
                                                          0.170***
## wash
##
                                (0.024)
                                                          (0.024)
##
                               -0.334***
                                                         -0.333***
## d1991
##
                                (0.024)
                                                          (0.024)
##
                               -0.343***
                                                         -0.347***
## d1992
##
                                (0.024)
                                                          (0.024)
##
                                                         -0.297***
## d1993
                               -0.296***
##
                                (0.023)
                                                          (0.023)
##
                               -0.247***
                                                         -0.248***
## d1994
##
                                (0.022)
                                                          (0.022)
##
```

```
## d1995
                            -0.144***
                                                   -0.146***
##
                            (0.022)
                                                    (0.022)
##
## d1996
                            -0.103***
                                                   -0.111***
##
                            (0.022)
                                                    (0.022)
##
                            -0.048**
                                                    -0.054**
## d1997
##
                            (0.022)
                                                    (0.022)
##
## d1998
                            0.089***
                                                    0.085***
##
                            (0.023)
                                                    (0.023)
##
                             0.038*
## d1999
                                                    0.033
##
                            (0.022)
                                                    (0.022)
##
## log(score):pinotnoir
                                                    -1.326**
                                                    (0.597)
                                                     -0.098
## log(score):cabernet
##
                                                    (0.577)
                                                   -2.613***
## log(score):merlot
##
                                                    (0.576)
## log(score):syrah
                                                   -2.610***
##
                                                    (0.691)
##
## Constant
                           -22.256***
                                                   -27.395***
                            (0.608)
                                                   (2.427)
## Observations 6,979
                                                    6,979
                           0.487
                                                    0.494
## Adjusted R2
                           0.486
                                                    0.492
## Residual Std. Error 0.378 (df = 6956) 0.376 (df = 6952)
## F Statistic 300.642*** (df = 22; 6956) 260.631*** (df = 26; 6952)
## -----
                            *p<0.1; **p<0.05; ***p<0.01
```

Q10)

```
# run linear hypothesis to test equivalency of effects
linearHypothesis(lm9_1, c('log(score):pinotnoir = log(score):cabernet', 'log(score):cabernet =
log(score):merlot', 'log(score):merlot = log(score):syrah'), vcov = vcovHC(lm9_1, 'HC1'))
## Linear hypothesis test
##
## Hypothesis:
## log(score):pinotnoir - log(score):cabernet = 0
## log(score):cabernet - log(score):merlot = 0
## log(score):merlot - log(score):syrah = 0
##
## Model 1: restricted model
## Model 2: log(price) ~ log(score) + log(score) * pinotnoir + log(score) *
##
       cabernet + log(score) * merlot + log(score) * syrah + pinotnoir +
##
       cabernet + merlot + syrah + napa + bayarea + sonoma + scoast +
##
       carn + sierra + mendo + wash + d1991 + d1992 + d1993 + d1994 +
       d1995 + d1996 + d1997 + d1998 + d1999
##
##
## Note: Coefficient covariance matrix supplied.
```

```
## Res.Df Df F Pr(>F)
## 1 6955
## 2 6952 3 26.781 < 2.2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```