assignment\_r

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Abstract

My solution to the Data analysis assignment in ØKA 201

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Been reading []Kivedal (2023)]

## Exercise 1:

price WinterRain temp HarvestRain   
 Min. : 495.2 Min. :376.0 Min. :14.98 Min. : 38.0   
 1st Qu.: 670.8 1st Qu.:543.5 1st Qu.:16.15 1st Qu.: 88.0   
 Median :1079.8 Median :600.0 Median :16.42 Median :123.0   
 Mean :1405.8 Mean :608.4 Mean :16.48 Mean :144.8   
 3rd Qu.:1707.7 3rd Qu.:705.5 3rd Qu.:17.01 3rd Qu.:185.5   
 Max. :4883.9 Max. :830.0 Max. :17.65 Max. :292.0   
 Age Dheavyraint   
 Min. : 3.00 Min. :0.0000   
 1st Qu.: 9.50 1st Qu.:0.0000   
 Median :16.00 Median :0.0000   
 Mean :16.19 Mean :0.1852   
 3rd Qu.:22.50 3rd Qu.:0.0000   
 Max. :31.00 Max. :1.0000

price WinterRain temp HarvestRain Age  
price 1.0000000 0.23384285 0.58888017 -0.44924408 0.45211288  
WinterRain 0.2338429 1.00000000 -0.32113230 -0.26798907 -0.05118354  
temp 0.5888802 -0.32113230 1.00000000 -0.02708361 0.29488335  
HarvestRain -0.4492441 -0.26798907 -0.02708361 1.00000000 0.05884976  
Age 0.4521129 -0.05118354 0.29488335 0.05884976 1.00000000  
Dheavyraint -0.3319703 -0.05574944 -0.03029117 0.81905472 -0.04625699  
 Dheavyraint  
price -0.33197026  
WinterRain -0.05574944  
temp -0.03029117  
HarvestRain 0.81905472  
Age -0.04625699  
Dheavyraint 1.00000000

## Exercise 2:

Call:  
lm(formula = for1, data = owine)  
  
Coefficients:  
(Intercept) Age   
 494.28 56.32

### a) Show output from Gretl

Model 3: OLS, using observations 1952-1980 (T = 27) Missing or incomplete observations dropped: 2 Dependent variable: price

|  | coefficent | std. error | t-ratio | p-value |
| --- | --- | --- | --- | --- |
| const | 494,278 | 402,112 | 1,229 | 0,2304 |
| Age | 56,3183 | 22,2218 | 2,534 | 0,0179 \*\* |

|  |  |  |  |
| --- | --- | --- | --- |
| Mean dependent var | 1405,800 | S.D. dependent var | 1027,226 |
| Sum squared resid | 21827151 | S.E. of regression | 934,3907 |
| R-squared | 0,204406 | Adjusted R-squared | 0,172582 |
| F(1, 25) | 6,423065 | P-value(F) | 0,017902 |
| Log-likelihood | −221,9495 | Akaike criterion | 447,8990 |
| Schwarz criterion | 450,4907 | Hannan-Quinn | 448,6697 |

### b) Interpret the estimated coefficients

494,278 is the estimated value of the dependent variable (price) when the age is zero.  
  
For every year the wine ages, the estimated price increases by 56,3183. We consider this statistically significant at the most commonly used 5% level of significance, since the p-value of 0.0171 is less than 0.05. This is also marked with two stars in the output from Gretl.

### c) Test whether the regressor has a significant effect on the dependent variable at a 1% level of significance

**Null Hypothesis (H₀)**: The Age coefficient = 0 (Age does not affect price).

**Alt. Hypothesis (H₁)**: The Age coefficient is not equal to 0 (Age does have an effect on price).

Testing on a 1% significance level with a p-value of 0.0179. The p-value is greater than 0.01, which means that we can’t reject the null hypothesis at the 1% significance level. In other words, we do not have enough statistical evidence that age has an effect on price at this level of significance.

### d) Make a prediction of the price when the independent variable is equal to its minimum value, mean value, and maximum value

The age is the independent variable. From the dataset we know that 3 is the minimum value, 16,185 is the mean value, and 31 is the maximum value. We also know that the constant is 494,278 and the coefficient for age is 56,3183.

**Minimum age:**

494.278 + (56.3183 \* 3)

[1] 663.2329

**Mean age:**

494.278 + (56.3183 \* 16.185)

[1] 1405.79

**Maximum age:**

494.278 + (56.3183 \* 31)

[1] 2240.145

### e) Calculate a 95% prediction interval for the price of the 1961 vintage

For 95% confidence intervals, t(25, 0,025) = 2,060

|  |  |  |  |  |
| --- | --- | --- | --- | --- |
|  | price | prediction | std. error | 95% interval |
| 1961 | 4883,90 | 1733,28 | 960,270 | -244,433 - 3710,99 |

Forecast evaluation statistics using 1 observations

|  |  |
| --- | --- |
| Mean Error | 3150,6 |
| Root Mean Squared Error | 3150,6 |
| Mean Absolute Error | 3150,6 |
| Mean Percentage Error | 64,51 |
| Mean Absolute Percentage Error | 64,51 |
| Theil’s U2 | 0 |

The 95% interval is [-244,433 - 3710,99]

## Exercise 3

Call:  
lm(formula = for2, data = owine)  
  
Coefficients:  
(Intercept) Age WinterRain temp HarvestRain   
 -15509.009 39.213 2.751 930.787 -5.047

### a) Show the output from Gretl

Model 1: OLS, using observations 1952-1980 (T = 27)  
Missing or incomplete observations dropped: 2  
Dependent variable: price

|  | coefficient | std. error | t-ratio | p-value |
| --- | --- | --- | --- | --- |
| const | -15509,0 | 3379,87 | -4,589 | 0,0001 \*\*\* |
| WinterRain | 2,75098 | 0,965119 | 2,850 | 0,0093 \*\*\* |
| Temp | 930,787 | 190,557 | 4,885 | 6,97e-05 \*\*\* |
| HarvestRain | -5,04694 | 1,61682 | -3,122 | 0,0050 \*\*\* |
| Age | 39,2126 | 14,3490 | 2,733 | 0,0121 \*\* |

|  |  |  |  |
| --- | --- | --- | --- |
| Mean dependent var | 1405,800 | S.D. dependent var | 1027,226 |
| Sum squared resid | 7238994 | S.E. of regression | 573,6246 |
| R-squared | 0,736141 | Adjusted R-squared | 0,688166 |
| F(4, 22) | 15,34443 | P-value(F) | 3,93e-06 |
| Log-likelihood | −207,0499 | Akaike criterion | 424,0999 |
| Schwarz criterion | 430,5791 | Hannan-Quinn | 426,0265 |

### b) Interpret the estimated slope coefficients

**WinterRain**: The coefficient is 2.75098, which means the price increases by that amount for each additional unit of WinterRain. We consider this statistically significant at the most commonly used 5% level of significance, since the p-value of 0.0093 is less than 0.05. The p-value of 0.0093 also means it’s statistically significant at the 1% level since it’s less than 0.001.  
  
**Temp**: The coefficient is 930.787, which means the price increases by that amount for each additional unit of Temp. We consider this statistically significant at the most commonly used 5% level of significance, since the p-value of 6.97e-05 is less than 0.05. The p-value of 6.97e-05 also means it’s statistically significant at the 1% level since it’s less than 0.001.  
  
**HarvestRain**:The coefficient is -5.04694, which means the price decreases by that amount for each additional unit of HarvestRain. We consider this statistically significant at the most commonly used 5% level of significance, since the p-value of 0.0050 is less than 0.05. The p-value of 0.0050 also means it’s statistically significant at the 1% level since it’s less than 0.001.  
  
**Age**: The coefficient is 39.2126, which means the price increases by that amount for each additional unit of Age. We consider this statistically significant at the most commonly used 5% level of significance, since the p-value of 0.0121 is less than 0.05. The p-value of 0.0121 also means it’s **not** statistically significant at the 1% level since it’s higher than 0.001.

### c) Compare with what you found in Exercise 2

Looking at age just like we did in exercise 2, we see that the coefficient is now 39,2126. In exercise 2, it was 56,3183. The difference is that in exercise 2, we didnt take WinterRain, Temp and HarvestRain into consideration. These other variables were unknown.  
  
We see now that the age increases less per unit compared to exercise 2 when we add the other variables to the equation. The difference is that we now estimate what effect one unit of age will have on the price, given that WinterRain, Temp and HarvestRain are the same.  
  
The p-value has decreased slightly, but not enough to push it below the 1% significance level.

## Exercise 4

Call:  
lm(formula = for3, data = owine)  
  
Coefficients:  
(Intercept) Age WinterRain temp HarvestRain Dheavyraint   
 -15289.642 39.916 2.671 923.652 -5.655 127.076

### a) Show the output from Gretl

Model 2: OLS, using observations 1952-1980 (T = 27)  
Missing or incomplete observations dropped: 2  
Dependent variable: price

|  | coefficient | std. error | t-ratio | p-value |
| --- | --- | --- | --- | --- |
| const | −13319,6 | 3910,53 | −3,406 | 0,0023  \*\*\* |
| Temp | 902,820 | 236,963 | 3,810 | 0,0009  \*\*\* |
| DUM | −815,924 | 394,606 | −2,068 | 0,0496  \*\* |

|  |  |  |  |
| --- | --- | --- | --- |
| Mean dependent var | 1405,800 | S.D. dependent var | 1027,226 |
| Sum squared resid | 15211368 | S.E. of regression | 796,120 |
| R-squared | 0,445550 | Adjusted R-squared | 0,399345 |
| F(2, 24) | 9,643056 | P-value(F) | 0,000844 |
| Log-likelihood | −217,0745 | Akaike criterion | 440,1490 |
| Schwarz criterion | 444,0365 | Hannan-Quinn | 441,3050 |

### b) Explain how high levels of rainfall in the harvest season can affect prices through fixed effects, and through interaction effects with temperature, based on the estimated model.

This model includes a dummy variable which is a fixed effect. In this case it’s used to capture the effect of rainfall above 200mm. When the rainfall is above this threshold, it predicts a decrease in the price with - 815,924.  
  
When it comes to the interaction with temperature, there is a possibility that high levels of rainfall means cooler temperatures. Seeing as a single unit of Temp estimates an increase in price of 902,820, we can say that higher temperatures are connected to higher prices. This means that if high rainfall affects the temperature in a negative way, it could lower the price.

## Exercise 5

### a) Calculate the residual for the 1961 vintage using Model 1, Model 2, and Model 3

The actual price of the 1961 vintage is 4,883.903, WinterRain is 830, temp is 17,3333, HarvestRain is 38, and Age is 22.  
  
  
**Model 1:**  
Price =

494.278 + (56.3183 \* 22)

[1] 1733.281

Residual for model 1 =

4883.903 - 1733.281

[1] 3150.622

**Model 2:**  
Price =

-15509.0+(2.75098\*830)+(930.787\*17.3333)+(-5.04694\*38)+(39.2126\*22)

[1] 3578.817

Residual for model 2 =

4883.903 - 3578.817

[1] 1305.086

**Model 3:**  
Price =

-13319.6+(902.820\*17.333)+(-815.924\*0)

[1] 2328.979

Residual for model 3 =

4883.903 - 2328.979

[1] 2554.924

### b) Comment on what you found in a), and this, in addition to other relevant information to discuss briefly which model you would prefer to use for predicting Bordeaux vintage wine prices

The residual for model 1 is quite large. This model only considers the age of the wine, and the model is probably too simple to correctly estimate the actual price of the 1961 vintage. The 1961 vintage is the most expensive out of all the wines, so the model probably needs more variables to correctly predict the price.  
  
The residual for model 2 is still large, but considerably less than for model 1. This model includes all the variables we have available. Still, it predicts a moderate price for the 1961 vintage, compared to the actual price. However, seeing as the 1961 vintage is such an outlier, it would probably be impossible to correctly predict the price.  
  
Model 3 faces the same challenges as model 1 and the residual is pretty large. There are not enough variables. It also mainly focuses on the outliers with high rainfall which are wines that usually are cheap. Another challenge with this model is that it doesn’t differentiate rainfalls between 0 and 200.  
  
Based on the residuals I would prefer model 2 for predicting Bordeaux vintage wine prices. While the residual is still big, it’s hard to predict outliers. Some outliers are even such anomalies that a model that predicts it right might be worse for predicting other Bordeax vintage wine prices. However, model 2 can still be improved by adding additional variables.

Conclusion:

**Content of** estm.R

R-code

```{r}  
## Internal  
library(OEKA201AssignmentMAT)  
  
## External  
library(readr)  
library(broom)  
library(gretlR)  
suppressPackageStartupMessages(library(dplyr))  
# Settings  
hlim <- 200  
# Reading data  
iwine <- wine  
owine <- iwine %>%  
 # variables in use  
 dplyr::select(price, WinterRain, temp, HarvestRain, Age) %>%  
 ## interaction effects (creates dummy variable)  
 dplyr::mutate(Dheavyraint=ifelse(HarvestRain>hlim,1,0)) %>%  
 ## na ommit  
 stats::na.omit()  
  
  
  
  
### Exercise 1  
sds <- summary(owine)  
cds <- cor(owine)  
### Exercise 2  
for1 <- price ~ Age   
mod1 <- lm(for1,data=owine)  
pv1 <- predict(mod1, newdata = owine)  
nd2 <- data.frame(Age = 1961)  
pv2 <- predict(mod1, newdata = owine, interval = "prediction", level = 0.95)  
  
### Exercise 3  
for2 <- price ~ Age + WinterRain + temp + HarvestRain   
mod2 <- lm(for2,data=owine)  
  
### Exercise 4  
for3 <- price ~ Age + WinterRain + temp + HarvestRain + Dheavyraint  
mod3 <- lm(for3,data=owine)  
  
### Exercise 5  
res1 <- resid(mod1)  
res2 <- resid(mod2)  
res3 <- resid(mod3)  
resf <- data.frame(res1,res2,res3)  
  
  
  
```

### References

Kivedal, Bjørnar Karlsen. 2023. *Anvendt Statistikk Og Økonometri*. "Universitetsforlaget".