

# EDS241: Assignment 4

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This question will ask you to estimate the price elasticity of demand for fresh sardines across 56 ports located in 4 European countries with monthly data from 2013 to 2019. The data are contained in the file EU\_sardines.csv, which is available on Gauchospace.

Each row in the data file is a combination of port location (where the fish is landed and sold) in a given year and month. You can ignore the fact that the sample is not balanced (the number of monthly observations varies across ports).

For the assignment, you will need the following variables: year, month, country, port (port where sardines are landed and sold), price\_euro\_kg (price per kg in €), and volume\_sold\_kg (quantity of sardines sold in kg). In the questions below, I use  $\log()$  to denote the natural logarithm.

**Read in the data:**

```
sardines <- read.csv(here("data", "EU_sardines.csv")) %>%  
  mutate(log_price_euro_kg = log(price_euro_kg), log_volume_sold_kg = log(volume_sold_kg))
```

**Question a:** Estimate a bivariate regression of  $\log(\text{volume\_sold\_kg})$  on  $\log(\text{price\_euro\_kg})$ . What is the price elasticity of demand for sardines? Test the null hypothesis that the price elasticity is equal to -1.

```
model_a <- lm_robust(data = sardines, log_volume_sold_kg ~ log_price_euro_kg)  
  
model_a_table <- broom::tidy(model_a) %>%  
  dplyr::select(term, estimate, std.error, p.value,) %>%  
  knitr::kable()  
  
model_a_table
```

term	estimate	std.error	p.value
(Intercept)	7.759061	0.0430246	0
log_price_euro_kg	-1.545335	0.0781254	0

```
linearHypothesis(model_a, c("log_price_euro_kg = -1"), white.adjust = "hc2")
```

```
## Linear hypothesis test  
##  
## Hypothesis:  
## log_price_euro_kg = - 1  
##
```

```
## Model 1: restricted model
## Model 2: log_volume_sold_kg ~ log_price_euro_kg
##
##      Res.Df Df    Chisq        Pr(>Chisq)
## 1      3987
## 2      3986   1 48.724 0.0000000000002946 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

The price elasticity of sardines is -1.55. Because our linear hypothesis gives us a very low p-value under 0.001, we can reject the null hypothesis that the price elasticity is equal to -1.

Question b: Like in Lecture 8 (see the IV.R script), we will use `wind_m_s` as an instrument for `log(price_euro_kg)`. To begin, estimate the first-stage regression relating `log(price_euro_kg)` to `wind_m_s`. Interpret the estimated coefficient on wind speed. Does it have the expected sign? Also test for the relevance of the instrument and whether it is a “weak” instrument by reporting the proper F-statistic.

```
model_b <- lm_robust(data = sardines, formula = log_price_euro_kg ~ wind_m_s)

model_b_table <- broom::tidy(model_b) %>%
  dplyr::select(term, estimate, std.error, p.value) %>%
  knitr::kable()

model_b_table
```

term	estimate	std.error	p.value
(Intercept)	-0.3048875	0.0273093	0
wind_m_s	0.0673459	0.0055995	0

```
summary(model_b)
```

[illegible]

The estimated coefficient for wind speed on  $\log(\text{price/kg})$  is 0.07, meaning that for every 1 m/s increase of wind speed, the  $\log(\text{price/kg})$  will increase by an estimated 0.07 euros. This is the sign that we would expect

to see because higher windspeed would make fishing more difficult, which would raise the price of fish. The F-statistic for wind speed is 144.6526487, so it is a relevant and non-weak instrument.

**Question c:** Estimate the TSLS estimator of the price elasticity of demand for sardines using `wind_m_s` as an instrument for `log(price_euro_kg)`. What is the estimated price elasticity of demand for sardines?

```
tsls <- ivreg(log_volume_sold_kg ~ log_price_euro_kg | wind_m_s, data = sardines)
summary(tsls)
```

```
##
## Call:
## ivreg(formula = log_volume_sold_kg ~ log_price_euro_kg | wind_m_s,
##       data = sardines)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -8.8626 -1.9790 -0.2333  2.0950  6.2354
##
## Coefficients:
##              Estimate Std. Error t value      Pr(>|t|)
## (Intercept)    7.75534    0.04331  179.08 <0.0000000000000002 ***
## log_price_euro_kg -1.08802    0.37003   -2.94      0.0033 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.728 on 3986 degrees of freedom
## Multiple R-Squared:  0.09529, Adjusted R-squared:  0.09506
## Wald test: 8.646 on 1 and 3986 DF, p-value: 0.003297
```

```
tsls_table <- broom::tidy(tsls) %>%
  dplyr::select(term, estimate, std.error, p.value) %>%
  knitr::kable()
```

```
tsls_table
```

term	estimate	std.error	p.value
(Intercept)	7.755341	0.0433068	0.0000000
log_price_euro_kg	-1.088015	0.3700295	0.0032973

The 2 stage least square estimator for price elasticity of demand for sardines, using wind speed as an instrument for `log(price_euro_kg)`, is -1.0880152.

**Question d:** Repeat the exercise in (c), but include fixed effects for each year, month, and country. [Hint: you can use the command “`as.factor(country) + as.factor(year) + as.factor(month)`” to the `ivreg` function in R]. Report the estimated price elasticity of demand and the F-statistic testing for relevant and non-weak instruments.

```
tsls_d <- ivreg(log_volume_sold_kg ~ log_price_euro_kg | wind_m_s +
  as.factor(country) +
  as.factor(year) +
  as.factor(month),
```

```

data = sardines)

summary(tsls_d)

##
## Call:
## ivreg(formula = log_volume_sold_kg ~ log_price_euro_kg | wind_m_s +
##       as.factor(country) + as.factor(year) + as.factor(month),
##       data = sardines)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -10.449  -2.153  -0.140   2.263   6.746
##
## Coefficients:
##              Estimate Std. Error t value      Pr(>|t|)
## (Intercept)    7.74350    0.04673 165.702 <0.0000000000000002 ***
## log_price_euro_kg 0.36697    0.31005   1.184      0.237
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.947 on 3986 degrees of freedom
## Multiple R-Squared:  -0.05549,    Adjusted R-squared:  -0.05576
## Wald test: 1.401 on 1 and 3986 DF,  p-value: 0.2366

tsls_d_table <- broom::tidy(tsls_d) %>%
  dplyr::select(term, estimate, std.error, p.value) %>%
  knitr::kable()

tsls_d_table

```

term	estimate	std.error	p.value
(Intercept)	7.7435045	0.0467316	0.0000000
log_price_euro_kg	0.3669746	0.3100514	0.2366449