

# EDS 241 Assignment 4

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## Introduction

We plan to estimate the price elasticity of demand for fresh sardines across 56 points in 4 European countries. We will use monthly data from 2013 - 2019.

## Data description

Variables include:

- country
- port: where the fish is landed and sold
- year
- month
- price\_euro\_kg: price per kg in Euros
- volume\_sold\_kg: quantity of sardines sold in kg
- wind\_m\_s

## Set up

### Read in the data

```
data <- read_csv(here("data", "EU_sardines.csv"))

data_clean <- data %>%
  clean_names() %>%
  mutate(year = as.factor(year),
         month = as.factor(month),
         price_euro_kg = as.numeric(price_euro_kg),
         volume_sold_kg = as.numeric(volume_sold_kg),
         wind_m_s = as.numeric(wind_m_s)
  )
```

(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.

```
data_log <- data_clean %>%
  mutate(volume_log = log(volume_sold_kg),
         price_log = log(price_euro_kg))

mdl <- lm_robust(price_log ~ volume_log, data = data_log)
price_elasticity <- round(mdl$coefficients[[2]], digits = 2)
```

The price elasticity of demand for sardines is approximately -0.07.

Table 1 shows the results of a bivariate regression of log-transformed volume sold and log-transformed price per kg of fresh sardines across 56 ports in 4 European countries from 2013 - 2019.

Table 1: Sardine volume significantly impacts price in Europe

<i>Dependent variable:</i>	
	Log(Price)
Log(Volume)	-0.068*** (0.003)
Observations	3,988
R <sup>2</sup>	0.104
<i>Note:</i> *p<0.1; **p<0.05; ***p<0.01	

(b) Like in Lecture 8 (see the IV.R script), we will use `wind_m_s` as an instrument for  $\log(\text{price\_euro\_kg})$ . To begin, estimate the first-stage regression relating  $\log(\text{price\_euro\_kg})$  to `wind_m_s`. Interpret the estimated coefficient on wind speed. Does it have the expected sign?

First-stage regression

```
fs1 <- lm(formula = price_log ~ wind_m_s, data=data_log)
summary(fs1)
```

```
##
## Call:
## lm(formula = price_log ~ wind_m_s, data = data_log)
##
## Residuals:
```

```
##      Min      1Q  Median      3Q      Max
## -4.4705 -0.3871 -0.0095  0.4060  3.7839
##
## Coefficients:
##              Estimate Std. Error t value      Pr(>|t|)
## (Intercept) -0.304888   0.026660  -11.44 <0.0000000000000002 ***
## wind_m_s      0.067346   0.005374   12.53 <0.0000000000000002 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.5884 on 3986 degrees of freedom
## Multiple R-squared:  0.0379, Adjusted R-squared:  0.03766
## F-statistic: 157 on 1 and 3986 DF, p-value: < 0.00000000000000022
```

```
wind_coef <- round(fs1$coefficients[[2]], digits = 2)
```

*#F > 10 BUT it is testing all parameters including the constant*

The coefficient on wind speed is approximately 0.07. In other words, on average, when wind speed increases by 1 m/sec, the log price of sardines per kg increases by 0.07 in Euros.

Table

(b) Also test for the relevance of the instrument and whether it is a “weak” instrument by reporting the proper F-statistic.

F-test for non-weak and relevant instruments

```
f <- linearHypothesis(fs1, c("wind_m_s=0"), white.adjust = "hc2")
```

```
f_coef <- round(f$F[2], digits = 2)
```

*#F > 10, which is above our threshold  
# p-value doesn't matter for us*

Wind speed is not a weak instrument because the F-test value of 144.65 is above our threshold of 10.

(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 - JUST-IDENTIFIED MODEL
# # Lecture 8, slide 13
# ts1s1 <- ivreg(log_tots ~ log_price | windspd, data = FULTON) #windspd = instrument
# summary(ts1s1)
```

```
# #takes cares of 1st and 2nd stage regression  
# #our elasticity of demand = B1 = -1.3744  
# #we estimate the demand elasticity (slope of the demand curve) using only the variation of log price  
# # 1 demand curve with multiple supply curves moving around (windspeed only affects supply, not demand.  
# #accounting for endogeneity bias can significantly impact our estimate of B1
```

(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.