

# Homework 3-2

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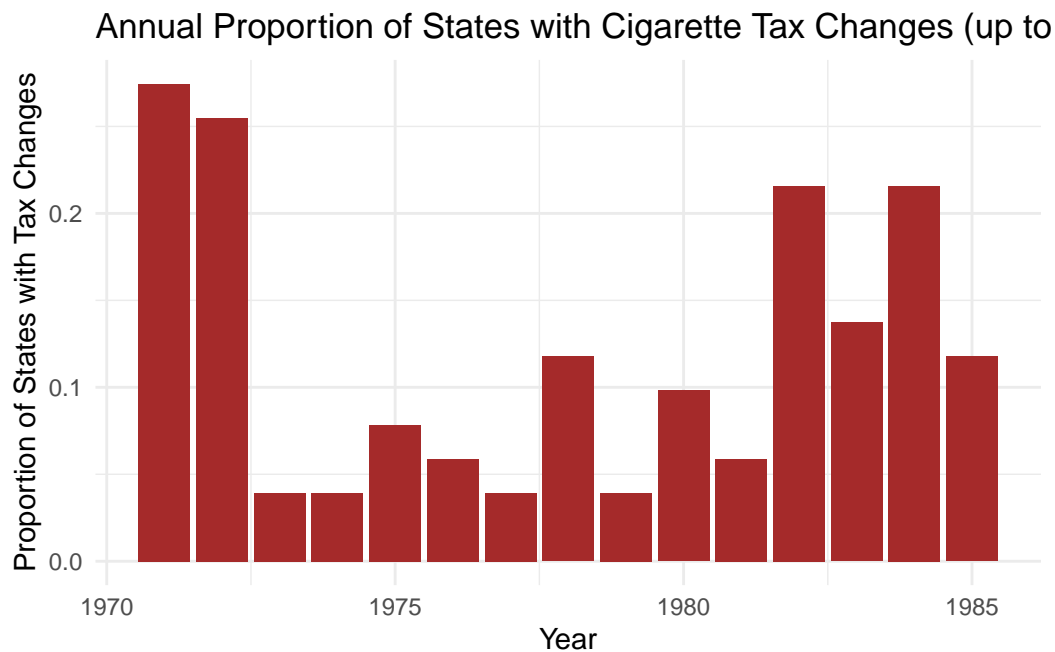
[Link to Github](#)

## Summarize The Data

1. Present a bar graph showing the proportion of states with a change in their cigarette tax in each year from 1970 to 1985.

```
yearlytaxchanges <- cig.data %>%  
  filter(Year <= 1985) %>%  
  group_by(Year) %>%  
  summarise(TaxChange = mean(tax_change_d, na.rm = TRUE))  
  
taxchangeplot <- ggplot(yearlytaxchanges, aes(x = Year, y = TaxChange)) +  
  geom_col(fill = "brown") +  
  labs(title = "Annual Proportion of States with Cigarette Tax Changes (up to 1985)",  
       x = "Year",  
       y = "Proportion of States with Tax Changes") +  
  theme_minimal()  
  
print(taxchangeplot)
```

Warning: Removed 1 rows containing missing values (`position\_stack()`).



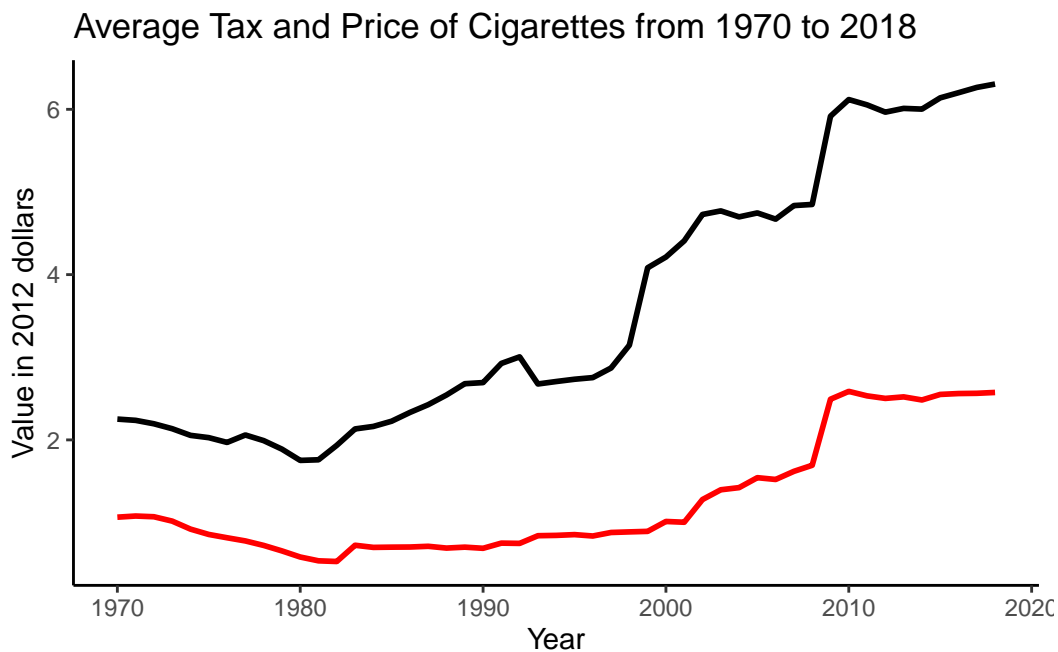
. Plot on a single graph the average tax (in 2012 dollars) on cigarettes and the average price of a pack of cigarettes from 1970 to 2018.

```
b.1970x2018 <- cig.data %>%
  filter(Year >= 1970 & Year <= 2018) %>%
  group_by(Year) %>%
  summarise(mean_price = mean(price_cpi_2022, na.rm = TRUE), mean_tax = mean(total_tax_cpi_2022, na.rm = TRUE))

b.1970x2018plot <- ggplot(b.1970x2018, aes(x = Year)) +
  geom_line(aes(y = mean_price), color = "black", size = 1) +
  geom_line(aes(y = mean_tax), color = "red", size = 1) +
  labs(title = "Average Tax and Price of Cigarettes from 1970 to 2018",
       x = "Year", y = "Value in 2012 dollars") +
  theme_classic()
```

Warning: Using `size` aesthetic for lines was deprecated in ggplot2 3.4.0.  
i Please use `linewidth` instead.

```
print(b.1970x2018plot)
```



. Identify the 5 states with the highest increases in cigarette prices (in dollars) over the time period. Plot the average number of packs sold per capita for those states from 1970 to 2018.

```
cigpack <- cig.data %>%
  filter(Year %in% c(1970, 2018)) %>%
  pivot_wider(names_from = Year, values_from = price_cpi_2022) %>%
  mutate(PriceIncrease = `2018` - `1970`) %>%
  arrange(desc(PriceIncrease)) %>%
  slice_head(n = 5)

top_states <- cigpack$state

print(top_states)
```

[1] "Alabama"	"Alabama"	"Alaska"
[4] "Alaska"	"Arizona"	"Arizona"
[7] "Arkansas"	"Arkansas"	"California"
[10] "California"	"Colorado"	"Colorado"
[13] "Connecticut"	"Connecticut"	"Delaware"
[16] "Delaware"	"District of Columbia"	"District of Columbia"
[19] "Florida"	"Florida"	"Georgia"
[22] "Georgia"	"Hawaii"	"Hawaii"
[25] "Idaho"	"Idaho"	"Illinois"
[28] "Illinois"	"Indiana"	"Indiana"
[31] "Iowa"	"Iowa"	"Kansas"
[34] "Kansas"	"Kentucky"	"Kentucky"
[37] "Louisiana"	"Louisiana"	"Maine"
[40] "Maine"	"Maryland"	"Maryland"
[43] "Massachusetts"	"Massachusetts"	"Michigan"
[46] "Michigan"	"Minnesota"	"Minnesota"
[49] "Mississippi"	"Mississippi"	"Missouri"
[52] "Missouri"	"Montana"	"Montana"
[55] "Nebraska"	"Nebraska"	"Nevada"
[58] "Nevada"	"New Hampshire"	"New Hampshire"
[61] "New Jersey"	"New Jersey"	"New Mexico"
[64] "New Mexico"	"New York"	"New York"
[67] "North Carolina"	"North Carolina"	"North Dakota"
[70] "North Dakota"	"Ohio"	"Ohio"
[73] "Oklahoma"	"Oklahoma"	"Oregon"
[76] "Oregon"	"Pennsylvania"	"Pennsylvania"
[79] "Rhode Island"	"Rhode Island"	"South Carolina"

```

[82] "South Carolina"      "South Dakota"      "South Dakota"
[85] "Tennessee"          "Tennessee"         "Texas"
[88] "Texas"               "Utah"              "Utah"
[91] "Vermont"             "Vermont"           "Virginia"
[94] "Virginia"            "Washington"        "Washington"
[97] "West Virginia"      "West Virginia"     "Wisconsin"
[100] "Wisconsin"           "Wyoming"           "Wyoming"

```

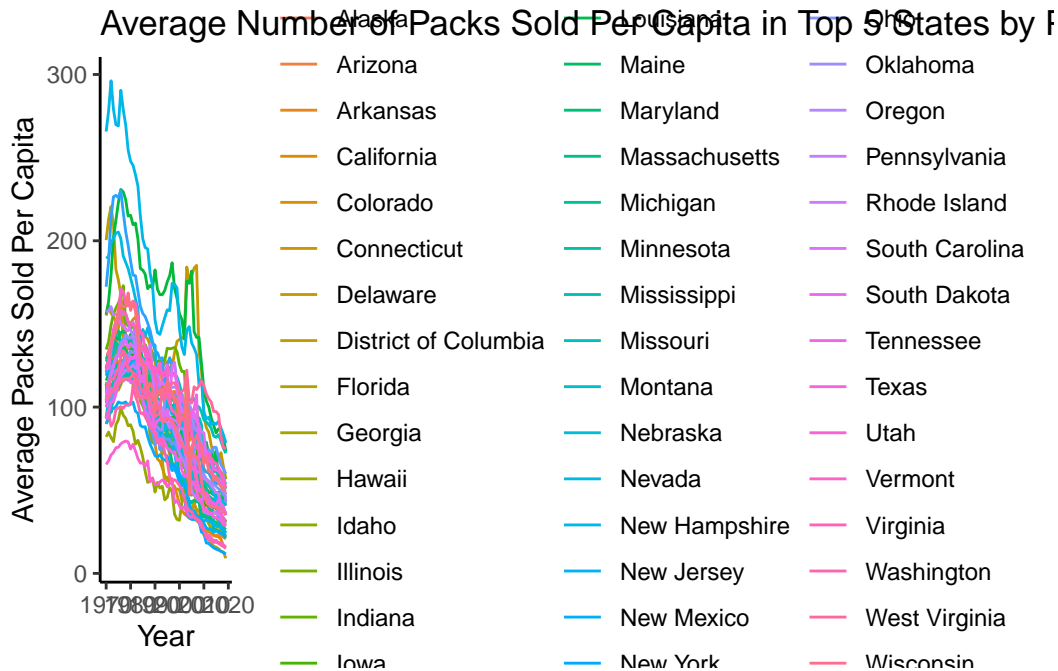
```

avg_sales_per_state <- cig.data %>%
  filter(state %in% top_states) %>%
  group_by(state, Year) %>%
  summarise(MeanSales = mean(sales_per_capita, na.rm = TRUE), .groups = 'drop')

avg_sales_plot <- ggplot(avg_sales_per_state, aes(x = Year, y = MeanSales, color = state)) +
  geom_line() +
  labs(title = "Average Number of Packs Sold Per Capita in Top 5 States by Price Increase fr",
       x = "Year",
       y = "Average Packs Sold Per Capita",
       color = "State") +
  theme_classic()

print(avg_sales_plot)

```



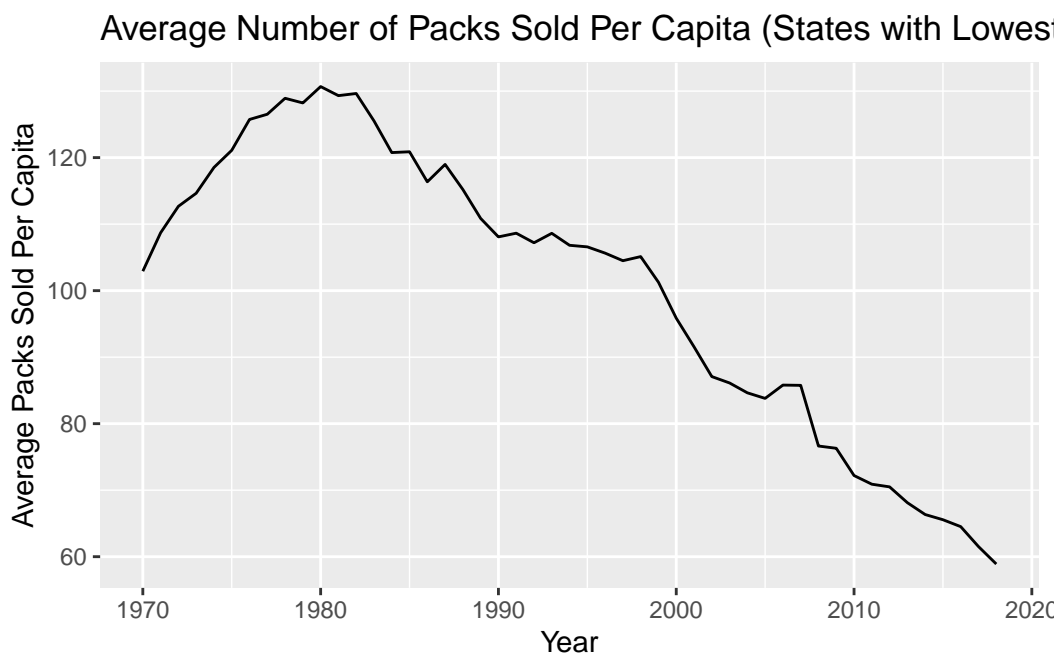
4. Identify the 5 states with the lowest increases in cigarette prices (in dollars) over the time period. Plot the average number of packs sold per capita for those states from 1970 to 2018.

```
bottom_states <- c("Missouri", "Tennessee", "North Dakota", "Alabama", "Georgia")

# calc n plot
avg_sales_bottom_states <- cig.data %>%
  filter(state %in% bottom_states, Year <= 2018) %>%
  group_by(Year) %>%
  summarise(MeanSales = mean(sales_per_capita, na.rm = TRUE))

avg_sales_bottom_plot <- ggplot(avg_sales_bottom_states, aes(x = Year, y = MeanSales)) +
  geom_line() +
  labs(title = "Average Number of Packs Sold Per Capita (States with Lowest Price Increase)",
       x = "Year",
       y = "Average Packs Sold Per Capita")

# Display the plot
print(avg_sales_bottom_plot)
```



. Compare the trends in sales from the 5 states with the highest price increases to those with the lowest price increases.

Both sets of groups showed a decreasing trend of cig sales over time. But the states with the higher price increases led to lesser sales than the ones with the lower increases.

## Estimate ATEs

Now let's work on estimating a demand curve for cigarettes. Specifically, we're going to estimate the price elasticity of demand for cigarettes. When explaining your findings, try to limit your discussion just to a couple of sentences.

6. Focusing only on the time period from 1970 to 1990, regress log sales on log prices to estimate the price elasticity of demand over that period. Interpret your results.

```
price_elasticity_model <- lm(ln_sales ~ ln_price_2012, data = filter(cig.data, Year >= 1970 & Year <= 1990))
model_summary <- summary(price_elasticity_model)
print(model_summary)
```

Call:

```
lm(formula = ln_sales ~ ln_price_2012, data = filter(cig.data,
  Year >= 1970 & Year <= 1990))
```

Residuals:

Min	1Q	Median	3Q	Max
-0.68335	-0.08598	-0.00284	0.08778	0.83516

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	5.42738	0.02975	182.4	<2e-16 ***
ln_price_2012	-0.80944	0.03837	-21.1	<2e-16 ***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.1894 on 1069 degrees of freedom

Multiple R-squared: 0.294, Adjusted R-squared: 0.2933

F-statistic: 445.1 on 1 and 1069 DF, p-value: < 2.2e-16

The coefficient is negative, which means as price goes up, sales go down and vice-versa.

7. Again limiting to 1970 to 1990, regress log sales on log prices using the total (federal and state) cigarette tax (in dollars) as an instrument for log prices. Interpret your results and compare your estimates to those without an instrument. Are they different? If so, why?



```
regdata <- cig.data %>%
  filter(Year >= 1970 & Year <= 1990)
summary(feols(ln_sales ~ 1 | ln_price_2012 ~ ln_tax_2012,
              data=regdata))
```

TSLS estimation, Dep. Var.: ln\_sales, Endo.: ln\_price\_2012, Instr.: ln\_tax\_2012

Second stage: Dep. Var.: ln\_sales

Observations: 1,071

Standard-errors: IID

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	5.416797	0.054494	99.4017	< 2.2e-16 ***
fit_ln_price_2012	-0.795524	0.071235	-11.1676	< 2.2e-16 ***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

RMSE: 0.189226 Adj. R2: 0.293235

F-test (1st stage), ln\_price\_2012: stat = 436.8 , p < 2.2e-16 , on 1 and 1,069 DoF.

Wu-Hausman: stat = 0.053709, p = 0.816775, on 1 and 1,068 DoF.

The significant negative coefficient for ln\_price\_2012 confirms that higher prices lead to lower sales. It doesn't seem like there's a significant difference with the estimates without the instrument (in this case the ln\_tax\_2012)

8. Show the first stage and reduced-form results from the instrument.

```
#q8
price_elasticity_model <- cig.data %>%
  filter(Year >= 1970 & Year <= 1990)

first_stage_model <- lm(ln_price_2012 ~ ln_tax_2012, data=price_elasticity_model)

price_elasticity_model$pricehat <- predict(first_stage_model)

second_stage_model <- lm(ln_sales ~ pricehat, data=price_elasticity_model)

summary(first_stage_model)
```

Call:

```
lm(formula = ln_price_2012 ~ ln_tax_2012, data = price_elasticity_model)
```

Residuals:

Min	1Q	Median	3Q	Max
-0.23046	-0.09207	-0.02919	0.08019	0.48675

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	0.839646	0.005421	154.9	<2e-16 ***
ln_tax_2012	0.260060	0.012443	20.9	<2e-16 ***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.1272 on 1069 degrees of freedom  
Multiple R-squared: 0.2901, Adjusted R-squared: 0.2894  
F-statistic: 436.8 on 1 and 1069 DF, p-value: < 2.2e-16

```
summary(second_stage_model)
```

Call:

```
lm(formula = ln_sales ~ pricehat, data = price_elasticity_model)
```

Residuals:

Min	1Q	Median	3Q	Max
-0.86239	-0.09798	0.00549	0.09359	0.95094

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	5.41680	0.06212	87.196	<2e-16 ***
pricehat	-0.79552	0.08121	-9.796	<2e-16 ***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.2159 on 1069 degrees of freedom  
Multiple R-squared: 0.08238, Adjusted R-squared: 0.08152  
F-statistic: 95.97 on 1 and 1069 DF, p-value: < 2.2e-16

```
reduced_form_model <- lm(ln_sales ~ ln_tax_2012, data=price_elasticity_model)
summary(reduced_form_model)
```

Call:

```
lm(formula = ln_sales ~ ln_tax_2012, data = price_elasticity_model)
```

Residuals:

Min	1Q	Median	3Q	Max
-0.86239	-0.09798	0.00549	0.09359	0.95094

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	4.748839	0.009202	516.092	<2e-16 ***
ln_tax_2012	-0.206884	0.021119	-9.796	<2e-16 ***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.2159 on 1069 degrees of freedom

Multiple R-squared: 0.08238, Adjusted R-squared: 0.08152

F-statistic: 95.97 on 1 and 1069 DF, p-value: < 2.2e-16

The coefficient for the second stage is -0.207 which is less an effect than the first stage result of -0.412. This is unexpected and may be a result of error as one might expect the decrease to be larger when accounting for the endogeneity of price.

9

. Repeat questions 1-3 focusing on the period from 1991 to 2015.

```
#q9
regdata2 <- cig.data %>%
filter(Year >= 1991 & Year <= 2015)

first_stage_model2 <- lm(ln_sales ~ ln_price_2012, data = regdata2)
summary(first_stage_model2)
```

Call:

```
lm(formula = ln_sales ~ ln_price_2012, data = regdata2)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-0.92230	-0.17004	0.00664	0.17869	1.10282

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	5.65996	0.03638	155.56	<2e-16 ***
ln_price_2012	-0.99681	0.02469	-40.37	<2e-16 ***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.296 on 1273 degrees of freedom

Multiple R-squared: 0.5614, Adjusted R-squared: 0.5611

F-statistic: 1630 on 1 and 1273 DF, p-value: < 2.2e-16

```
summary(feols(ln_sales ~ 1 | ln_tax_2012,
              data= regdata2))
```

OLS estimation, Dep. Var.: ln\_sales

Observations: 1,275

Fixed-effects: ln\_tax\_2012: 1,024

RMSE: 0.112223 Adj. R2: 0.679548

```
first_stage2 <- lm(ln_price_2012 ~ ln_tax_2012, data = regdata2)
summary(first_stage2)
```

```

Call:
lm(formula = ln_price_2012 ~ ln_tax_2012, data = regdata2)

Residuals:
    Min       1Q   Median       3Q      Max
-0.36750 -0.09020  0.00725  0.08241  0.45045

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)  1.315073   0.004386  299.84  <2e-16 ***
ln_tax_2012  0.513550   0.006922   74.19  <2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.1456 on 1273 degrees of freedom
Multiple R-squared:  0.8121,    Adjusted R-squared:  0.812
F-statistic: 5504 on 1 and 1273 DF,  p-value: < 2.2e-16

reduced_form_model2 <- lm(ln_sales ~ ln_tax_2012, data = regdata2)
summary(reduced_form_model2)

```

```

Call:
lm(formula = ln_sales ~ ln_tax_2012, data = regdata2)

Residuals:
    Min       1Q   Median       3Q      Max
-0.90878 -0.15465  0.01119  0.15334  1.16925

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)  4.36742   0.00844  517.50  <2e-16 ***
ln_tax_2012 -0.59063   0.01332  -44.34  <2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.2802 on 1273 degrees of freedom
Multiple R-squared:  0.607, Adjusted R-squared:  0.6067
F-statistic: 1966 on 1 and 1273 DF,  p-value: < 2.2e-16

```

10. Compare your elasticity estimates from 1970-1990 versus those from 1991-2015. Are they different? If so, why?.

For the years 1991-2015, the elasticity estimate of -0.99681 mean there is a stronger reaction of cigarette sales to price changes than in the earlier period of 1970-1990, which had an elasticity estimate of -0.795524. It seems that as time went on, people became more sensitive to the price of cigarettes. This shift could be due to several factors—perhaps there's been a change in how society views smoking, or maybe there are now more alternatives available to smokers.