

Subjective evaluation of active noise cancellation in headphones

```
In [ ]: import numpy as np
```

Dummy Data Generation

```
In [ ]: import random as rnd

samples_quantity = 30
measurements_per_sample = 8
```

$X_1 \equiv$ ANC Status

```
In [ ]: anc_status = np.empty((samples_quantity, measurements_per_sample))

for i in range(samples_quantity):
    for j in range(0, measurements_per_sample, 4):
        anc_status[i][j] = 0 # Sin ANC
        anc_status[i][j+1] = 0 # Sin ANC
        anc_status[i][j+2] = 1 # Con ANC
        anc_status[i][j+3] = 1 # Con ANC
```

$X_2 \equiv$ Measured Noise

```
In [ ]: import math

measured_noise_spl = np.empty((samples_quantity, measurements_per_sample))

noise_mean = 65
noise_deviation = 2
background_noise_spl = rnd.gauss(noise_mean, noise_deviation)
delta_spl = 15
generated_noise_spl = noise_mean + 3 * noise_deviation + delta_spl

for i in range(samples_quantity):
    for j in range(measurements_per_sample):
        background_noise_spl = rnd.gauss(noise_mean, noise_deviation)

        if (j % 2 == 0):
            measured_noise_spl[i][j] = background_noise_spl # Solo ruido
        else:
            measured_noise_spl[i][j] = 10 * math.log10(10 ** (background_
```

$Y_1, Y_2 \equiv$ Perceived Noise Level and Sound Quality

```
In [ ]: perceived_noise = np.empty((samples_quantity, measurements_per_sample))
perceived_quality = np.empty((samples_quantity, measurements_per_sample))
```

```

for i in range(samples_quantity):
    for j in range(0, measurements_per_sample, 4):
        perceived_noise[i][j] = rnd.randint(1, 4) # Sin ANC, poco ruido
        perceived_noise[i][j+1] = rnd.randint(4, 5) # Sin ANC, mucho ruido
        perceived_noise[i][j+2] = rnd.randint(1, 2) # Con ANC, poco ruido
        perceived_noise[i][j+3] = rnd.randint(1, 3) # Con ANC, mucho ruido

        perceived_quality[i][j] = rnd.randint(3, 5) # Sin ANC, poco ruido
        perceived_quality[i][j+1] = rnd.randint(3, 5) # Sin ANC, mucho ruido
        perceived_quality[i][j+2] = rnd.randint(2, 5) # Con ANC, poco ruido
        perceived_quality[i][j+3] = rnd.randint(1, 4) # Con ANC, mucho ruido

```

Results Dot Plots

```

In [ ]: import seaborn as sns
import matplotlib.pyplot as plt

x_1 = anc_status.ravel()
x_2 = measured_noise_spl.ravel()
y_1 = perceived_noise.ravel()
y_2 = perceived_quality.ravel()

```

$X_2 \equiv$ Measured Noise Level Dot Plot

```

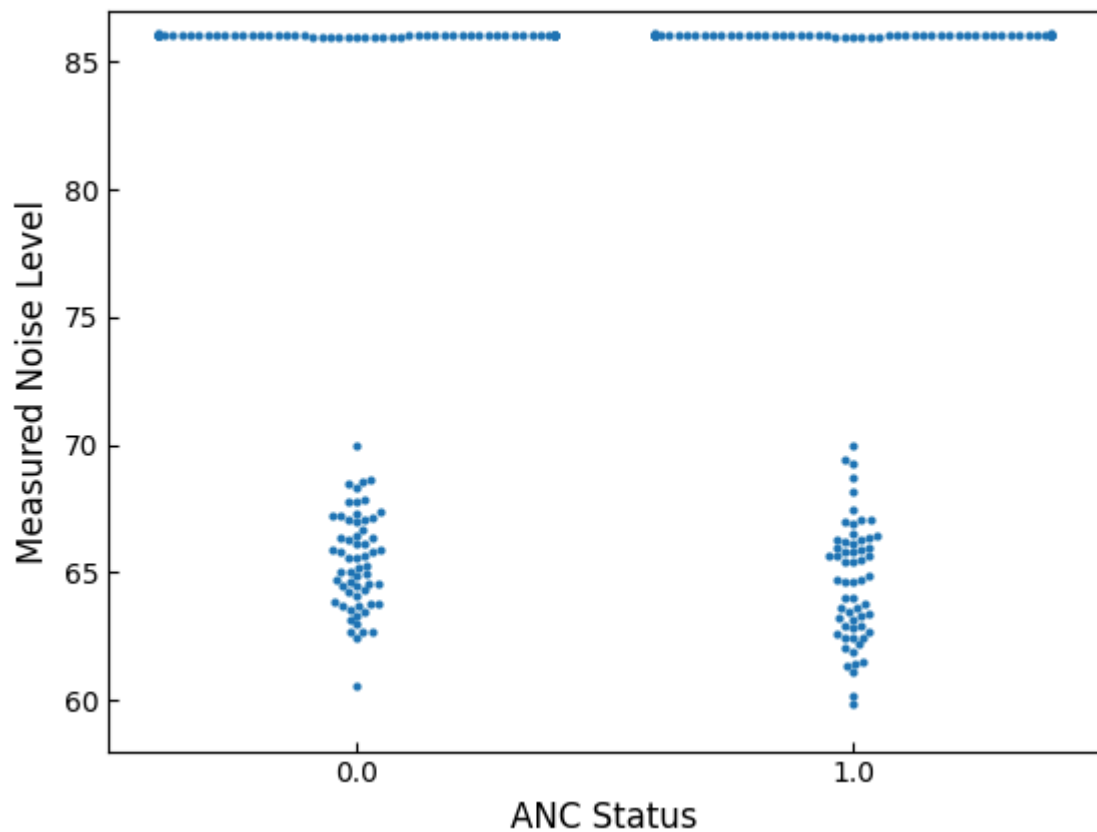
In [ ]: y_min = noise_mean - 3 * noise_deviation - 1
y_max = noise_mean + 3 * noise_deviation + delta_spl + 1

sns.swarmplot(x = x_1, y = x_2, size = 3)
plt.tick_params(axis = 'both', direction = 'in')
plt.ylim(y_min, y_max)
plt.xlabel('ANC Status', fontsize = 12)
plt.ylabel('Measured Noise Level', fontsize = 12)
plt.show()

```

/Users/mrmalvicino/Desktop/GitHub/anc/myenv/lib/python3.12/site-packages/seaborn/categorical.py:3399: UserWarning: 12.5% of the points cannot be placed; you may want to decrease the size of the markers or use stripplot.

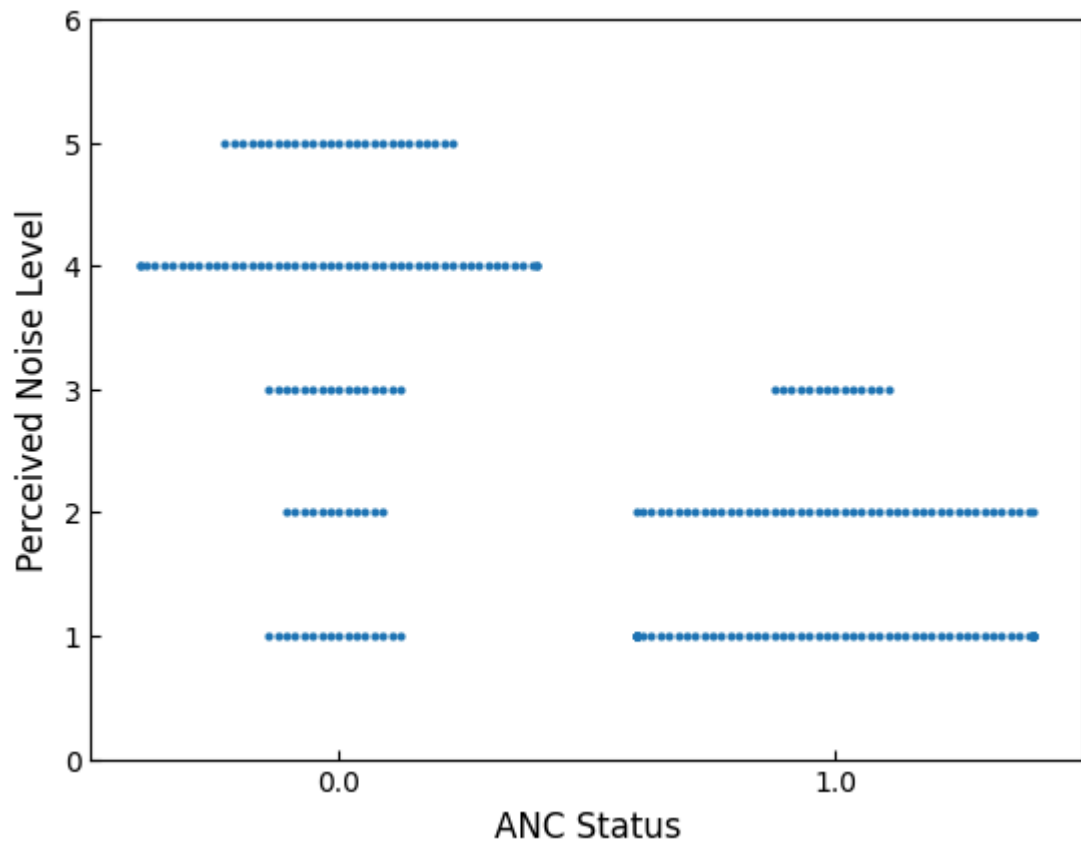
warnings.warn(msg, UserWarning)



$Y_1 \equiv$ Perceived Noise Level Dot Plot

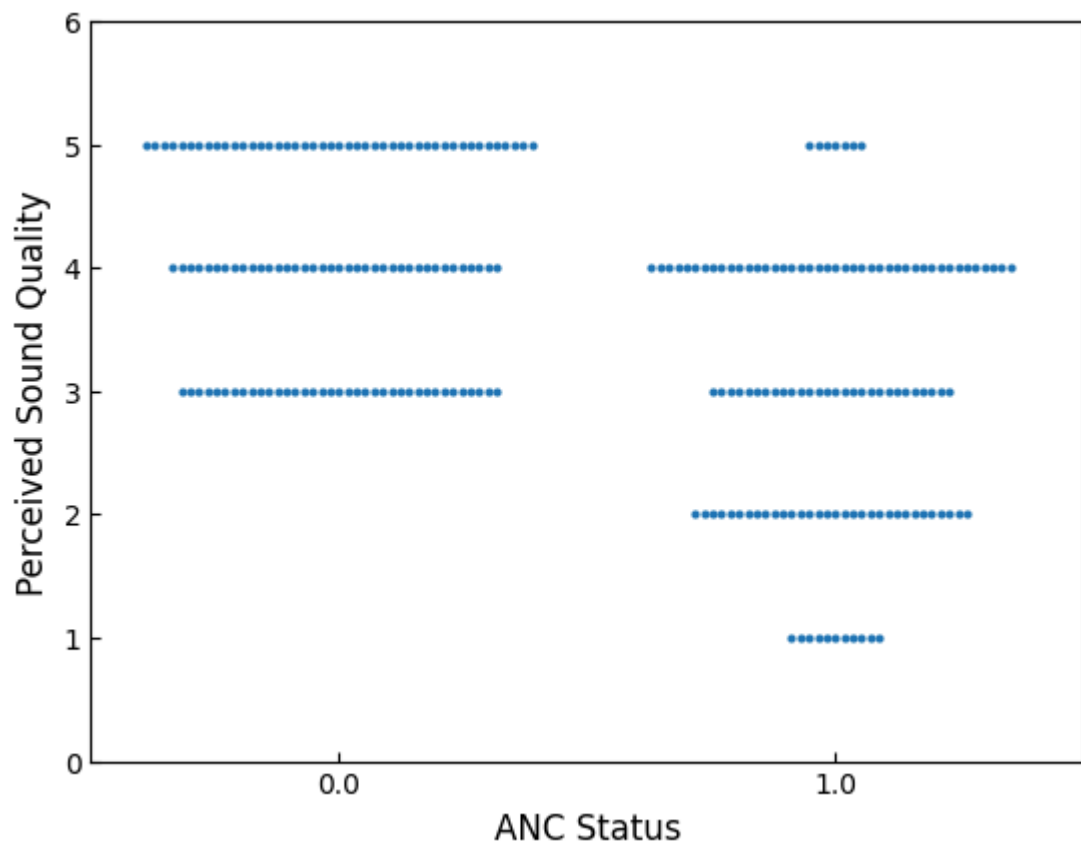
```
In [ ]: sns.swarmplot(x = x_1, y = y_1, size = 3)
plt.tick_params(axis = 'both', direction = 'in')
plt.ylim(0, 6)
plt.xlabel('ANC Status', fontsize = 12)
plt.ylabel('Perceived Noise Level', fontsize = 12)
plt.show()
```

/Users/mrmalvicino/Desktop/GitHub/anc/myenv/lib/python3.12/site-packages/seaborn/categorical.py:3399: UserWarning: 13.3% of the points cannot be placed; you may want to decrease the size of the markers or use stripplot.
warnings.warn(msg, UserWarning)



$Y_2 \equiv$ Perceived Sound Quality Dot Plot

```
In [ ]: sns.swarmplot(x = x_1, y = y_2, size = 3)
plt.tick_params(axis = 'both', direction = 'in')
plt.ylim(0, 6)
plt.xlabel('ANC Status', fontsize = 12)
plt.ylabel('Perceived Sound Quality', fontsize = 12)
plt.show()
```



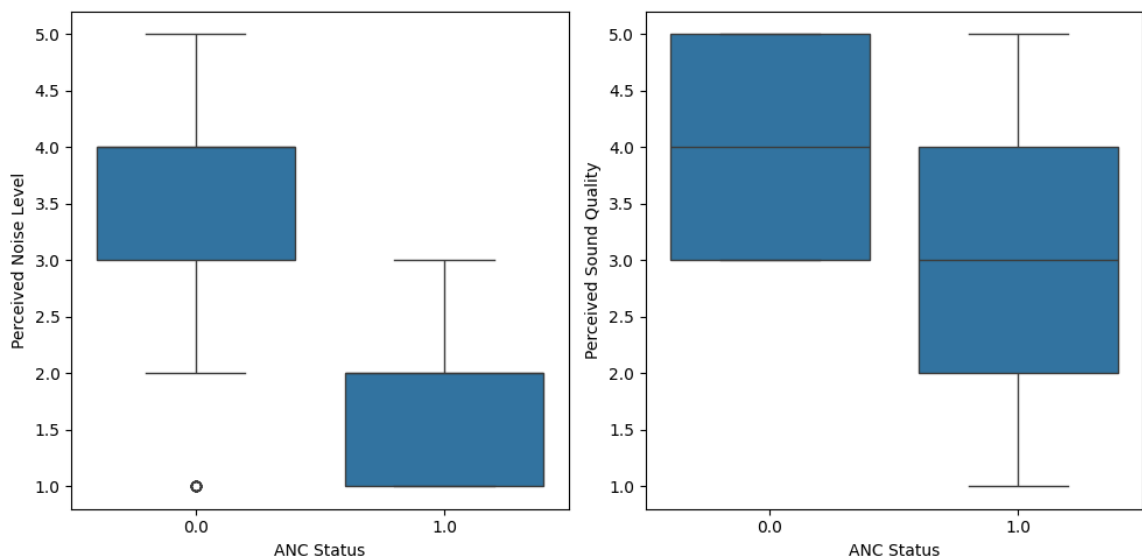
Data Validation

Box Plot

```
In [ ]: plt.figure(figsize=(10, 5))
plt.subplot(1, 2, 1)
sns.boxplot(x = x_1, y = y_1)
plt.xlabel('ANC Status')
plt.ylabel('Perceived Noise Level')

plt.subplot(1, 2, 2)
sns.boxplot(x = x_1, y = y_2)
plt.xlabel('ANC Status')
plt.ylabel('Perceived Sound Quality')

plt.tight_layout()
plt.show()
```



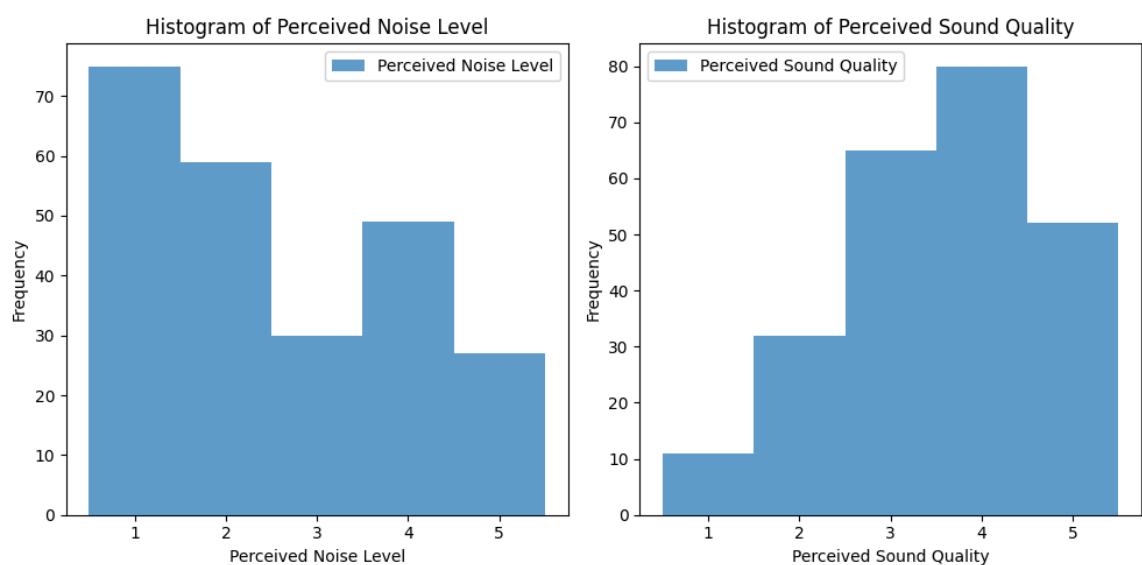
Histogram

```
In [ ]: plt.figure(figsize=(10, 5))

plt.subplot(1, 2, 1)
plt.hist(y_1, bins=np.arange(0.5, 6.5), alpha=0.7, label='Perceived Noise Level')
plt.xlabel('Perceived Noise Level')
plt.ylabel('Frequency')
plt.title('Histogram of Perceived Noise Level')
plt.legend()

plt.subplot(1, 2, 2)
plt.hist(y_2, bins=np.arange(0.5, 6.5), alpha=0.7, label='Perceived Sound Quality')
plt.xlabel('Perceived Sound Quality')
plt.ylabel('Frequency')
plt.title('Histogram of Perceived Sound Quality')
plt.legend()

plt.tight_layout()
plt.show()
```



Shapiro-Wilk Normality test

```
In [ ]: from scipy import stats

statistic, p_value = stats.shapiro(x_2)

print("X2:")
print(f"Statistic: {statistic:.3f}")
print(f"p-value: {p_value:.3f}")

statistic, p_value = stats.shapiro(y_1)

print("\nY1:")
print(f"Statistic: {statistic:.3f}")
print(f"p-value: {p_value:.3f}")

statistic, p_value = stats.shapiro(y_2)

print("\nY2:")
print(f"Statistic: {statistic:.3f}")
print(f"p-value: {p_value:.3f}")
```

X2:
Statistic: 0.720
p-value: 0.000

Y1:
Statistic: 0.859
p-value: 0.000

Y2:
Statistic: 0.896
p-value: 0.000

Correlation and Regression

```
In [ ]: import statsmodels.api as sm
```

Scatter Plot and Pearson Correlation for $Y_1(X_1)$

```
In [ ]: # plt.scatter(x_1, y_1)
# plt.tick_params(axis = 'both', direction = 'in')
# plt.xlim(-1, 2)
# plt.ylim(0, 6)
# plt.xlabel('Variable X1 [ANC Status]', fontsize = 12)
# plt.ylabel('Variable Y1 [Noise score]', fontsize = 12)
# plt.show()

correlation_coefficient, p_value = stats.pearsonr(x_1, y_1)

print(f"r = {correlation_coefficient:.3f}")
print(f"p-value = {p_value:.3f}")

r = -0.668
p-value = 0.000
```

Scatter Plot and Pearson Correlation for $Y_2(X_1)$

```
In [ ]: # plt.scatter(x_1, y_2)
# plt.tick_params(axis = 'both', direction = 'in')
# plt.xlim(-1, 2)
# plt.ylim(0, 6)
# plt.xlabel('Variable X1 [ANC Status]', fontsize = 12)
# plt.ylabel('Variable Y2 [Quality score]', fontsize = 12)
# plt.show()

correlation_coefficient, p_value = stats.pearsonr(x_1, y_2)

print(f"r = {correlation_coefficient:.3f}")
print(f"p-value = {p_value:.3f}")

r = -0.475
p-value = 0.000
```

Linear Regression for $Y_1(X_1)$

```
In [ ]: x_1_c = sm.add_constant(x_1)
model = sm.OLS(y_1, x_1_c).fit()
print(model.summary())
predictions = model.predict(x_1_c)

plt.scatter(x_1, y_1, label='Datos observados')
plt.plot(x_1, predictions, color='red', label='Línea de regresión')
plt.xlim(-0.5, 1.5)
plt.ylim(0, 6)
plt.xlabel('ANC Status')
plt.ylabel('Perceived Quality')
plt.legend()
plt.show()
```

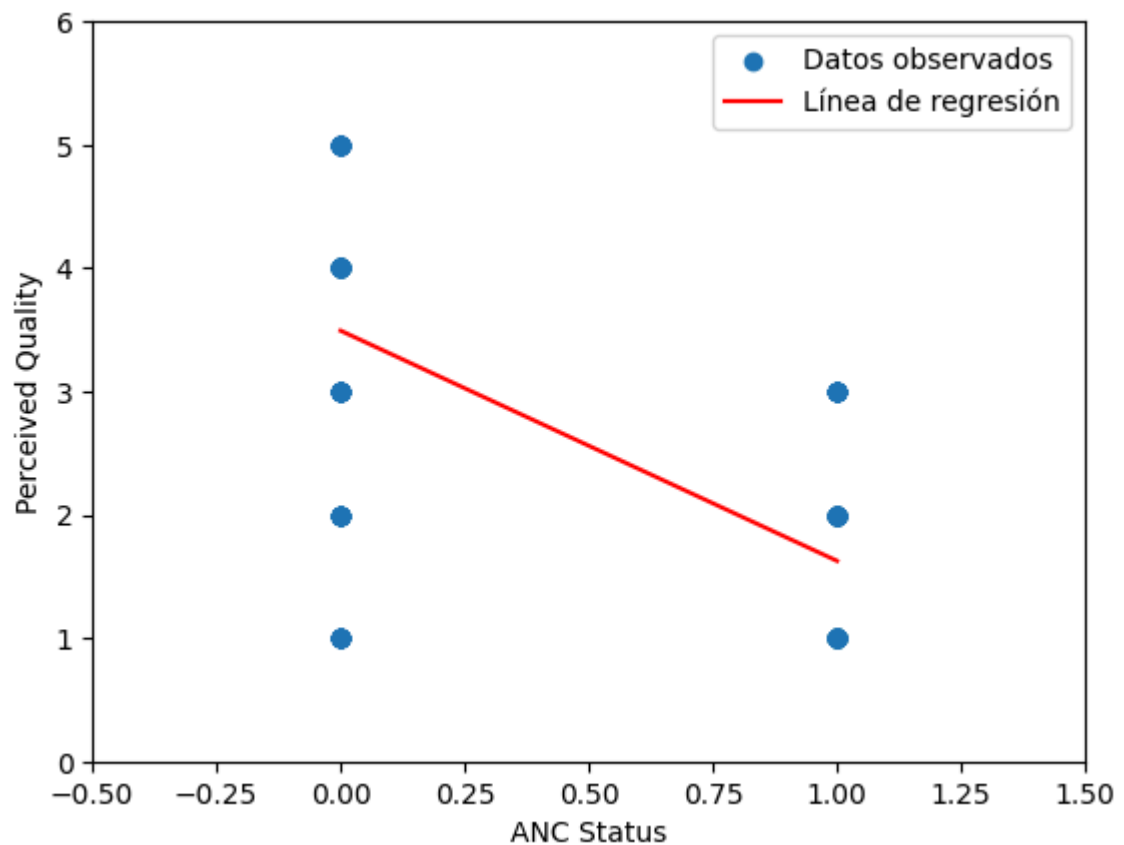

OLS Regression Results

=====					
Dep. Variable: y R-squared: 0.446					
Model: OLS Adj. R-squared: 0.443					
Method: Least Squares F-statistic: 191.3					
Date: Fri, 07 Jun 2024 Prob (F-statistic): 2.52e-32					
Time: 20:09:19 Log-Likelihood: -350.20					
No. Observations: 240 AIC: 704.4					
Df Residuals: 238 BIC: 711.4					
Df Model: 1					
Covariance Type: nonrobust					
=====					
=====					
	coef	std err	t	P> t	[0.025
0.975]					

const	3.4917	0.095	36.587	0.000	3.304
3.680					
x1	-1.8667	0.135	-13.831	0.000	-2.133
-1.601					
=====					
=====					
Omnibus: 14.070 Durbin-Watson: 2.644					
Prob(Omnibus): 0.001 Jarque-Bera (JB): 15.121					
Skew: -0.612 Prob(JB): 0.000521					
Kurtosis: 3.122 Cond. No. 2.62					
=====					
=====					

Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.



Linear Regression for $Y_2(X_1)$

```
In [ ]: x_1_c = sm.add_constant(x_1)
model = sm.OLS(y_2, x_1_c).fit()
print(model.summary())
predictions = model.predict(x_1_c)

plt.scatter(x_1, y_2, label='Datos observados')
plt.plot(x_1, predictions, color='red', label='Línea de regresión')
plt.xlim(-0.5, 1.5)
plt.ylim(0, 6)
plt.xlabel('ANC Status')
plt.ylabel('Perceived Quality')
plt.legend()
plt.show()
```

OLS Regression Results

=====					
Dep. Variable:		y	R-squared:		
0.225					
Model:		OLS	Adj. R-squared:		
0.222					
Method:		Least Squares	F-statistic:		
69.22					
Date:		Fri, 07 Jun 2024	Prob (F-statistic):		6.
86e-15					
Time:		20:09:19	Log-Likelihood:		-
334.09					
No. Observations:		240	AIC:		
672.2					
Df Residuals:		238	BIC:		
679.1					
Df Model:		1			
Covariance Type:		nonrobust			
=====					
=====					
	coef	std err	t	P> t	[0.025
0.975]					

const	4.0667	0.089	45.571	0.000	3.891
4.242					
x1	-1.0500	0.126	-8.320	0.000	-1.299
-0.801					
=====					
=====					
Omnibus:		32.242	Durbin-Watson:		
2.236					
Prob(Omnibus):		0.000	Jarque-Bera (JB):		
9.837					
Skew:		-0.170	Prob(JB):		
0.00731					
Kurtosis:		2.068	Cond. No.		
2.62					
=====					
=====					

Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

