Actividad: Regresión Lineal 2

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Regresión lineal múltiple

Utiliza un modelo de regresión lineal múltiple para predecir el radio del tumor. Las variables regresoras de tu modelo deben de ser todas las variables de la base de datos.

```
In [1]:
        # Cargamos las librerias necesarias para la actividad
        import pandas as pd
        import numpy as np
        import statsmodels.api as sm
        from sklearn.model_selection import train test split
        import statsmodels.formula.api as smf
        from sklearn.preprocessing import StandardScaler
        from sklearn.preprocessing import LabelEncoder
        import seaborn as sns
        import matplotlib.pyplot as plt
        from scipy.stats import t
        import scipy.stats as stats
        df = pd.read csv('breast cancer.csv')
        df.drop(['id','diagnosis','concave points mean','concave points_se','concave points_worst'],axis
        =1,inplace=True)
```

1.- Base de datos completa. No se observan valores faltantes. En caso de haberlos se realiza imputación simple.

```
In [2]:
        df.isnull().sum()
Out[2]:
        radius mean
                                    0
        texture_mean
        perimeter_mean
        area mean
        smoothness_mean
        compactness mean
                                    0
        concavity_mean
                                    0
        symmetry mean
                                    0
        fractal dimension mean
        radius_se
                                    0
        texture se
                                    0
        perimeter_se
                                    0
        area_se
        smoothness se
        compactness_se
        concavity se
        symmetry_se
        fractal_dimension_se
        radius worst
                                    0
                                    0
        texture_worst
        perimeter worst
                                    0
        area_worst
        smoothness worst
        compactness worst
                                    0
        concavity_worst
        symmetry worst
        fractal_dimension_worst
        dtype: int64
```

2.-Mostrar que las variables regresoras son independientes. En caso de no serlo realizar el procedimiento correspondiente.

En la matriz de correlación se aprecia mejor entre que variables se esta presentando tanto bajas como altas correlaciones.

Como los datos mostraron alta correlación fue necesario hacer una estandarización de los datos

```
In [7]: #Entrenamiento y prueba del modelo
    entrenamiento, prueba = train_test_split(df_estandar, test_size=0.2, random_state=42)

#Modelo OLS
modelo = smf.ols(formula='radius_mean~texture_mean+perimeter_mean+area_mean+smoothness_mean+comp
    actness_mean+concavity_mean+symmetry_mean+fractal_dimension_mean+radius_se+texture_se+perimeter_
    se+area_se+smoothness_se+compactness_se+concavity_se+symmetry_se+fractal_dimension_se+radius_wor
    st+texture_worst+perimeter_worst+area_worst+smoothness_worst+compactness_worst+concavity_worst+s
    ymmetry_worst+fractal_dimension_worst', data=entrenamiento)
    modelo = modelo.fit()
    print(modelo.summary())
```

OLS Regression Results

============	:==========	=======================================	==========
Dep. Variable:	radius_mean	R-squared:	1.000
Model:	OLS	Adj. R-squared:	1.000
Method:	Least Squares	F-statistic:	6.611e+04
Date:	Mon, 04 Sep 2023	<pre>Prob (F-statistic):</pre>	0.00
Time:	23:18:24	Log-Likelihood:	1240.8
No. Observations:	455	AIC:	-2428.
Df Residuals:	428	BIC:	-2316.
- 6			

Df Model: 26

Covariance Type: nonrobust

	coef	std err	t 	P> t	[0.025	0.975]
Intercept	0.0005	0.001	0.630	0.529	-0.001	0.002
texture_mean	-0.0016	0.003	-0.598	0.550	-0.007	0.004
perimeter_mean	0.9492	0.018	54.007	0.000	0.915	0.984
area_mean	0.0715	0.013	5.299	0.000	0.045	0.098
smoothness_mean	0.0067	0.002	3.253	0.001	0.003	0.011
compactness_mean	-0.0565	0.005	-11.860	0.000	-0.066	-0.047
concavity_mean	-0.0363	0.004	-8.830	0.000	-0.044	-0.028
symmetry_mean	0.0038	0.002	2.443	0.015	0.001	0.007
<pre>fractal_dimension_mean</pre>	0.0072	0.003	2.382	0.018	0.001	0.013
radius_se	0.0045	0.006	0.694	0.488	-0.008	0.017
texture_se	-9.373e-05	0.002	-0.058	0.953	-0.003	0.003
perimeter_se	-0.0163	0.006	-2.742	0.006	-0.028	-0.005
area_se	0.0006	0.004	0.129	0.897	-0.008	0.009
smoothness_se	0.0014	0.001	0.958	0.338	-0.001	0.004
compactness_se	-0.0018	0.003	-0.662	0.508	-0.007	0.004
concavity_se	0.0144	0.002	6.440	0.000	0.010	0.019
symmetry_se	0.0044	0.002	2.462	0.014	0.001	0.008
<pre>fractal_dimension_se</pre>	-0.0032	0.002	-1.415	0.158	-0.008	0.001
radius_worst	0.2323	0.018	12.784	0.000	0.197	0.268
texture_worst	0.0002	0.003	0.059	0.953	-0.006	0.007
perimeter_worst	-0.1139	0.015	-7.626	0.000	-0.143	-0.085
area_worst	-0.0840	0.013	-6.369	0.000	-0.110	-0.058
smoothness_worst	-0.0049	0.002	-2.064	0.040	-0.010	-0.000
compactness_worst	0.0157	0.005	3.477	0.001	0.007	0.025
concavity_worst	0.0010	0.004	0.268	0.788	-0.007	0.009

```
symmetry worst
                           -0.0048
                                         0.002
                                                   -2.069
                                                               0.039
                                                                           -0.009
                                                                                       -0.000
fractal dimension worst
                           -0.0035
                                         0.003
                                                   -1.050
                                                               0.294
                                                                           -0.010
                                                                                        0.003
Omnibus:
                                46.518
                                         Durbin-Watson:
                                                                           2.076
Prob(Omnibus):
                                0.000
                                         Jarque-Bera (JB):
                                                                         200.782
Skew:
                                0.306
                                         Prob(JB):
                                                                        2.52e-44
Kurtosis:
                                6.196
                                         Cond. No.
                                                                            120.
```

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

Haciendo un análisis del anterior modelo, el valor del estadístico F es extremadamente alto y la probabilidad asociada cercana a cero indican que el modelo en su conjunto es estadísticamente significativo. Esto sugiere que al menos una de las variables independientes incluidas en el modelo tiene un impacto significativo en la variable dependiente (radius_mean).

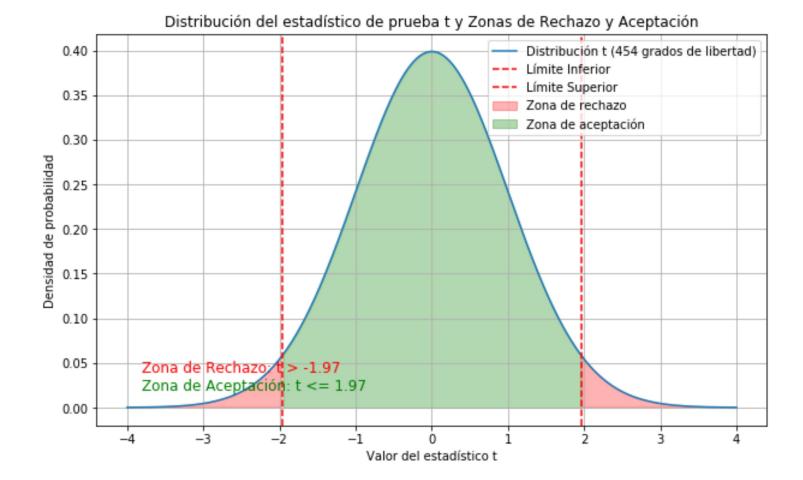
3.-Hipótesis nula de los coeficientes de regresión. Estadístico de prueba, distribución del estadístico de prueba.

Para un 95% de confianza realiza un diagrama en donde se muestre la distribución del estadístico de prueba, la zona de aceptación y la zona de rechazo.

```
In [8]: # Calcular los estadísticos t
    nivel_de_confianza = 0.95
    alpha2 = (1 - nivel_de_confianza) / 2
    grados_de_libertad = len(entrenamiento) - 1
    valor_critico_t = t.ppf(1 - alpha2, df=grados_de_libertad)

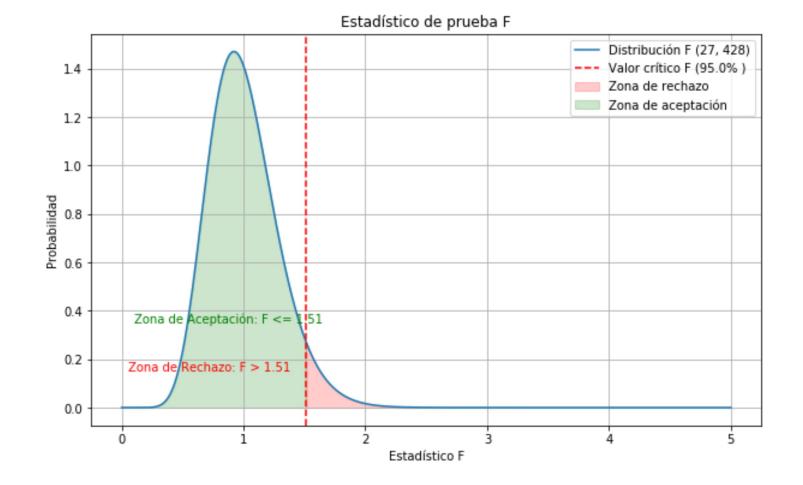
limite_inferior = -valor_critico_t
    limite_superior = valor_critico_t
```

```
In [9]: # Rango de valores para el estadístico t
        rango_t = np.linspace(-4, 4, 400)
        densidad_t = t.pdf(rango_t, df=grados_de_libertad) # Distribución t-Student
        # Crear el gráfico
        plt.figure(figsize=(10, 6))
        plt.plot(rango_t, densidad_t, label=f'Distribución t ({grados_de_libertad} grados de libertad)')
        plt.fill between(rango t, 0, densidad t, where=(rango t < limite inferior) | (rango t > limite s
        uperior), color='red', alpha=0.3, label='Zona de rechazo')
        plt.fill between(rango t, 0, densidad t, where=(rango t >= limite inferior) & (rango t <= limite
        superior), color='green', alpha=0.3, label='Zona de aceptación')
        # Agrega líneas verticales para el estadístico de prueba y los límites
        plt.axvline(limite inferior, color='red', linestyle='--', label='Límite Inferior')
        plt.axvline(limite superior, color='red', linestyle='--', label='Límite Superior')
        # Etiquetas y Leyenda
        plt.title('Distribución del estadístico de prueba t y Zonas de Rechazo y Aceptación')
        plt.xlabel('Valor del estadístico t')
        plt.ylabel('Densidad de probabilidad')
        plt.legend()
        plt.grid()
        # Etiquetas en el gráfico
        plt.text(-3.8, 0.04, f'Zona de Rechazo: t > {limite_inferior:.2f}', fontsize=12, color='red')
        plt.text(-3.8, 0.02, f'Zona de Aceptación: t <= {limite superior:.2f}', fontsize=12, color='gree
        n')
        # Mostrar el gráfico
        plt.show()
```



4.-Hipótesis nula de la significancia del modelo (prueba F-Fisher). Menciona que distribución tiene el estadístico de prueba con qué número de grados de libertad. Para un 95% de confianza realiza un diagrama en donde se muestre la distribución del estadístico de prueba, la zona de aceptación y la zona de rechazo.

```
In [10]: from scipy.stats import f
         coeficientes = modelo.params
         # Grados de libertad del modelo y del error
         df model = len(coeficientes)
         df error = len(entrenamiento) - len(coeficientes)
         # Nivel de confianza
         nivel de confianza = 0.95
         # Valor crítico F para el nivel de confianza y grados de libertad
         valor critico F = f.ppf(nivel de confianza, dfn=df model, dfd=df error)
         rango F = np.linspace(0, 5, 1000)
         densidad F = f.pdf(rango F, dfn=df model, dfd=df error)
         plt.figure(figsize=(10, 6))
         plt.plot(rango_F, densidad_F, label=f'Distribución F ({df_model}, {df_error})')
         plt.axvline(x=valor critico F, color='red', linestyle='--', label=f'Valor crítico F ({nivel de c
         onfianza * 100}% )')
         plt.fill between(rango F, densidad F, where=((rango F > valor critico F)), color='red', alpha=0.
         2, label='Zona de rechazo')
         plt.fill_between(rango_F, densidad_F, where=((rango_F <= valor_critico_F)), color='green', alpha
         =0.2, label='Zona de aceptación')
         plt.title('Estadístico de prueba F')
         plt.xlabel('Estadístico F')
         plt.ylabel('Probabilidad')
         plt.legend()
         plt.grid()
         plt.text(0.05, 0.15, f'Zona de Rechazo: F > {valor critico F:.2f}', fontsize=10, color='red')
         plt.text(0.1, 0.35, f'Zona de Aceptación: F <= {valor_critico_F:.2f}', fontsize=10, color='green
          ')
         plt.show()
```



5.- Realiza un modelo de regresión hacia atrás (backward). Explica el criterio para ir eliminando variables del modelo.

Explicando un poco mejor el modelo de regresión backward, en pocas palabras, es una técnica utilizada para simplificar un modelo de regresión múltiple eliminando gradualmente las variables predictoras que tienen un impacto menos significativo en la predicción de la variable dependiente. Su significancia se evalúa si su p-valor es mayor a 0.05.

En cada paso, se eliminará la variable independiente menos significativa, y el modelo se ajustará nuevamente sin esa variable.

```
In [11]: # Tomamos las variables para X y Y en nuestro modelo de regresion backward
         df estandar['intercept'] = 1
         X = df estandar[[ 'texture mean', 'perimeter mean', 'area mean', 'smoothness mean',
              'compactness mean', 'concavity mean', 'symmetry mean', 'fractal dimension mean',
              'radius_se', 'perimeter_se', 'area_se', 'smoothness_se',
              'compactness se', 'concavity se', 'symmetry se', 'fractal dimension se',
              'radius worst', 'texture worst', 'perimeter worst', 'area worst',
              'smoothness worst', 'compactness worst', 'concavity worst', 'symmetry worst',
              'fractal dimension worst']]
         Y = df estandar['radius mean']
         modelo2 = sm.OLS(Y, X).fit()
         while any(modelo2.pvalues > 0.05):
             variable menos significativa = modelo2.pvalues.idxmax()
             X = X.drop(variable menos significativa, axis=1)
             modelo2 = sm.OLS(Y, X).fit()
             print('\n--- La variable eliminada fue: ',variable menos significativa)
             print('\n')
             print(modelo2.summary())
```

OLS Regression Results

Dep. Variable: radius mean R-squared (uncentered): 1.000 Model: OLS Adj. R-squared (uncentered): 1.000 Method: Least Squares F-statistic: 8.357e+04 Date: Mon, 04 Sep 2023 Prob (F-statistic): 0.00 Time: 23:18:25 Log-Likelihood: 1528.7 No. Observations: 569 AIC: -3009. Df Residuals: BIC: -2905. 545

Df Model: 24

Covariance Type: nonrobust

=======================================	:=======:	========	=======	========	========	========
	coef	std err	t	P> t	[0.025	0.975]
texture_mean	-0.0012	0.001	-1.443	0.150	-0.003	0.000
perimeter_mean	0.9379	0.016	60.464	0.000	0.907	0.968
area_mean	0.0797	0.012	6.675	0.000	0.056	0.103
smoothness_mean	0.0088	0.002	4.746	0.000	0.005	0.012
compactness_mean	-0.0570	0.004	-13.277	0.000	-0.065	-0.049
concavity_mean	-0.0341	0.004	-8.922	0.000	-0.042	-0.027
symmetry_mean	0.0029	0.001	2.020	0.044	7.87e-05	0.006
<pre>fractal_dimension_mean</pre>	0.0037	0.003	1.333	0.183	-0.002	0.009
radius_se	0.0093	0.006	1.573	0.116	-0.002	0.021
perimeter_se	-0.0191	0.005	-3.506	0.000	-0.030	-0.008
area_se	-0.0013	0.004	-0.308	0.759	-0.009	0.007
smoothness_se	0.0025	0.001	1.818	0.070	-0.000	0.005
compactness_se	-0.0015	0.003	-0.570	0.569	-0.007	0.004
concavity_se	0.0139	0.002	6.578	0.000	0.010	0.018
symmetry_se	0.0005	0.001	0.346	0.730	-0.002	0.003
<pre>fractal_dimension_se</pre>	-0.0018	0.002	-0.837	0.403	-0.006	0.002
radius_worst	0.2181	0.017	12.823	0.000	0.185	0.251
perimeter_worst	-0.0985	0.013	-7.379	0.000	-0.125	-0.072
area_worst	-0.0839	0.012	-7.208	0.000	-0.107	-0.061
smoothness_worst	-0.0073	0.002	-3.417	0.001	-0.012	-0.003
compactness_worst	0.0158	0.004	3.783	0.000	0.008	0.024
concavity_worst	-0.0004	0.004	-0.106	0.916	-0.008	0.007

symmetry_worst	-0.0027	0.002 -1.304	4 0.193	-0.007	0.001	
<pre>fractal_dimension_worst</pre>	-0.0025	0.003 -0.804	4 0.422	-0.008	0.004	
=======================================	=========	============	========	=======		
Omnibus:	76.188	Durbin-Watson:		1.916		
Prob(Omnibus):	0.000	Jarque-Bera (JB):	691.023		
Skew:	-0.135	<pre>Prob(JB):</pre>		8.84e-151		
Kurtosis:	8.392	Cond. No. 116.		116.		

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

--- La variable eliminada fue: concavity_worst

	OLS R	Regression	n Results			
Dep. Variable: Model: Method: Date: Time: No. Observations: Df Residuals: Df Model: Covariance Type:	radius_mean OLS Least Squares Mon, 04 Sep 2023 23:18:25 569 546 23 nonrobust	Adj. R F-stat: Prob (I	red (uncenter -squared (unc istic: statistic): kelihood:	entered):	8.7	1.000 1.000 737e+04 0.00 1528.7 -3011.
=======================================	coef	std err	 t	P> t	[0.025	0.975]
texture_mean perimeter_mean area_mean smoothness_mean compactness_mean concavity_mean symmetry_mean fractal_dimension_mea radius_se perimeter_se	-0.0012 0.9378 0.0797 0.0088 -0.0569 -0.0343 0.0029 n 0.0037 0.0094 -0.0192	0.001 0.015 0.012 0.002 0.004 0.003 0.001 0.003 0.006 0.005	-1.444 60.544 6.709 4.759 -13.736 -11.349 2.020 1.350 1.583 -3.512	0.149 0.000 0.000 0.000 0.000 0.044 0.177 0.114 0.000	-0.003 0.907 0.056 0.005 -0.065 -0.040 7.93e-05 -0.002 -0.002	0.000 0.968 0.103 0.012 -0.049 -0.028 0.006 0.009 0.021 -0.008
area_se	-0.0013	0.004	-0.306	0.759	-0.009	0.007

smoothness_se	0.0025	0.001	1.843	0.066	-0.000	0.005
compactness_se	-0.0015	0.003	-0.563	0.574	-0.007	0.004
concavity_se	0.0138	0.002	7.480	0.000	0.010	0.017
symmetry_se	0.0005	0.001	0.341	0.733	-0.002	0.003
<pre>fractal_dimension_se</pre>	-0.0018	0.002	-0.835	0.404	-0.006	0.002
radius_worst	0.2180	0.017	12.834	0.000	0.185	0.251
perimeter_worst	-0.0985	0.013	-7.386	0.000	-0.125	-0.072
area_worst	-0.0840	0.012	-7.218	0.000	-0.107	-0.061
smoothness_worst	-0.0074	0.002	-3.460	0.001	-0.012	-0.003
compactness_worst	0.0156	0.004	4.166	0.000	0.008	0.023
symmetry_worst	-0.0027	0.002	-1.302	0.194	-0.007	0.001
<pre>fractal_dimension_worst</pre>	-0.0025	0.003	-0.851	0.395	-0.008	0.003
Omnibus:	76.132	:====== Durbin-۱	======= Watson:	=======	1.917	
Prob(Omnibus):	0.000		Bera (JB):	689.166		
Skew:	-0.136	Prob(JB	` '	2	.24e-150	
Kurtosis:	8.385	Cond. No	•		112.	
=======================================	=========	=======:	========	=======	======	

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

--- La variable eliminada fue: area_se

Dep. Variable:	radius_mean	R-squar	ed (uncenter	ed):		1.000
Model:	OLS	Adj. R-	squared (unc	entered):		1.000
Method:	Least Squares	F-stati	stic:		9.14	l9e+04
Date:	Mon, 04 Sep 2023	Prob (F	<pre>-statistic):</pre>			0.00
Time:	23:18:25	Log-Lik	elihood:		1	.528.6
No. Observations:	569	AIC:			-	3013.
Df Residuals:	547	BIC:			-	2918.
Df Model:	22					
Covariance Type:	nonrobust					
=======================================	coef	std err	t	P> t	[0.025	0.975]
texture_mean	-0.0011	0.001	-1.426	0.155	-0.003	0.000

perimeter_mean	0.9363	0.015	64.087	0.000	0.908	0.965
area_mean	0.0803	0.012	6.853	0.000	0.057	0.103
smoothness_mean	0.0088	0.002	4.768	0.000	0.005	0.012
compactness_mean	-0.0568	0.004	-13.760	0.000	-0.065	-0.049
concavity_mean	-0.0343	0.003	-11.354	0.000	-0.040	-0.028
symmetry_mean	0.0029	0.001	2.119	0.035	0.000	0.006
<pre>fractal_dimension_mean</pre>	0.0038	0.003	1.395	0.164	-0.002	0.009
radius_se	0.0086	0.005	1.619	0.106	-0.002	0.019
perimeter_se	-0.0195	0.005	-3.678	0.000	-0.030	-0.009
smoothness_se	0.0025	0.001	1.888	0.060	-0.000	0.005
compactness_se	-0.0015	0.003	-0.565	0.572	-0.007	0.004
concavity_se	0.0138	0.002	7.482	0.000	0.010	0.017
symmetry_se	0.0006	0.001	0.425	0.671	-0.002	0.003
<pre>fractal_dimension_se</pre>	-0.0017	0.002	-0.816	0.415	-0.006	0.002
radius_worst	0.2201	0.016	14.187	0.000	0.190	0.251
perimeter_worst	-0.0977	0.013	-7.474	0.000	-0.123	-0.072
area_worst	-0.0860	0.010	-9.016	0.000	-0.105	-0.067
smoothness_worst	-0.0074	0.002	-3.516	0.000	-0.012	-0.003
compactness_worst	0.0155	0.004	4.161	0.000	0.008	0.023
symmetry_worst	-0.0028	0.002	-1.415	0.158	-0.007	0.001
fractal_dimension_worst	-0.0026	0.003	-0.884	0.377	-0.008	0.003
Omnibus:	75.876	 -Durbin	Watson:		1.917	
Prob(Omnibus):	0.000	Jarque-	Bera (JB):		685.394	
Skew:	-0.132	Prob(JB):	1	.47e-149	
Kurtosis:	8.370	Cond. N	0.		99.0	
		•				

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

--- La variable eliminada fue: symmetry_se

===========			
Dep. Variable:	radius_mean	R-squared (uncentered):	1.000
Model:	OLS	Adj. R-squared (uncentered):	1.000
Method:	Least Squares	F-statistic:	9.599e+04
Date:	Mon, 04 Sep 2023	<pre>Prob (F-statistic):</pre>	0.00

Time:	23:18:25	Log-Lik	elihood:			1528.5
No. Observations:	569	AIC:				-3015.
Df Residuals:	548	BIC:				-2924.
Df Model:	21					
Covariance Type:	nonrobust					
	coef	std err	t	P> t	[0.025	0.975]
texture_mean	-0.0011	0.001	-1.393	0.164	-0.003	0.000
perimeter_mean	0.9369	0.015	64.490	0.000	0.908	0.965
area_mean	0.0800	0.012	6.845	0.000	0.057	0.103
smoothness_mean	0.0089	0.002	4.952	0.000	0.005	0.012
compactness_mean	-0.0568	0.004	-13.773	0.000	-0.065	-0.049
concavity_mean	-0.0342	0.003	-11.356	0.000	-0.040	-0.028
symmetry_mean	0.0028	0.001	2.084	0.038	0.000	0.005
<pre>fractal_dimension_mean</pre>	0.0039	0.003	1.451	0.147	-0.001	0.009
radius_se	0.0086	0.005	1.617	0.106	-0.002	0.019
perimeter_se	-0.0193	0.005	-3.656	0.000	-0.030	-0.009
smoothness_se	0.0027	0.001	2.183	0.029	0.000	0.005
compactness_se	-0.0014	0.003	-0.515	0.607	-0.007	0.004
concavity_se	0.0138	0.002	7.479	0.000	0.010	0.017
<pre>fractal_dimension_se</pre>	-0.0017	0.002	-0.792	0.429	-0.006	0.002
radius_worst	0.2199	0.015	14.192	0.000	0.189	0.250
perimeter_worst	-0.0983	0.013	-7.568	0.000	-0.124	-0.073
area_worst	-0.0857	0.010	-9.015	0.000	-0.104	-0.067
smoothness_worst	-0.0077	0.002	-3.897	0.000	-0.012	-0.004
compactness_worst	0.0154	0.004	4.144	0.000	0.008	0.023
symmetry_worst	-0.0022	0.001	-1.604	0.109	-0.005	0.000
<pre>fractal_dimension_worst</pre>	-0.0028	0.003	-0.969	0.333	-0.009	0.003
=======================================	=========	=======	========	========	======	
Omnibus:	73.545				1.919	
Prob(Omnibus):		•	Bera (JB):		641.568	
Skew:		Prob(JB	•	4	.85e-140	
Kurtosis:	8.197	Cond. N	lo.		98.6	
	=========	=======	========	=======	======	

Warnings:

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

⁻⁻⁻ La variable eliminada fue: compactness_se

Dep. Variable:	radius_mear	•	ed (uncenter		1.000		
Model:			Adj. R-squared (uncentered):			1.000	
Method:	Least Squares				1.00	9e+05	
	lon, 04 Sep 2023	•	-statistic):		_	0.00	
Time:	23:18:25	•	elihood:			528.4	
No. Observations:	569					3017.	
Df Residuals:	549				-	2930.	
Df Model:	26						
Covariance Type:	nonrobust 	: :=======	========	:=======	========	=======	
	coef	std err	t	P> t	[0.025	0.975	
texture_mean	-0.0011	0.001	-1.392	0.165	-0.003	0.000	
perimeter_mean	0.9364	0.014	64.618	0.000	0.908	0.96	
area_mean	0.0804	0.012	6.890	0.000	0.057	0.10	
smoothness_mean	0.0090	0.002	4.969	0.000	0.005	0.01	
compactness_mean	-0.0573	0.004	-14.181	0.000	-0.065	-0.049	
concavity_mean	-0.0339	0.003	-11.533	0.000	-0.040	-0.02	
symmetry_mean	0.0028	0.001	2.094	0.037	0.000	0.00	
fractal_dimension_mear	0.0040	0.003	1.487	0.138	-0.001	0.00	
radius_se	0.0090	0.005	1.718	0.086	-0.001	0.019	
perimeter_se	-0.0198	0.005	-3.832	0.000	-0.030	-0.010	
smoothness_se	0.0025	0.001	2.139	0.033	0.000	0.00	
concavity_se	0.0134	0.002	7.912	0.000	0.010	0.01	
<pre>fractal_dimension_se</pre>	-0.0023	0.002	-1.329	0.184	-0.006	0.00	
radius_worst	0.2197	0.015	14.192	0.000	0.189	0.25	
perimeter_worst	-0.0973	0.013	-7.582	0.000	-0.123	-0.072	
area_worst	-0.0862	0.009	-9.116	0.000	-0.105	-0.068	
smoothness_worst	-0.0075	0.002	-3.873	0.000	-0.011	-0.004	
compactness_worst	0.0144	0.003	4.589	0.000	0.008	0.023	
symmetry_worst	-0.0022	0.001	-1.608	0.108	-0.005	0.000	
fractal_dimension_wors	t -0.0024	0.003	-0.852	0.395	-0.008	0.003	

Omnibus: 72.564 Durbin-Watson: 1.922 Prob(Omnibus): 0.000 Jarque-Bera (JB): 622.855 Skew: -0.101 Prob(JB): 5.61e-136

Kurtosis: 8.122 Cond. No.	96.4
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[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

--- La variable eliminada fue: fractal_dimension_worst

OLS Regression Results

===========	:=========:::	=======================================	=========
Dep. Variable:	radius_mean	R-squared (uncentered):	1.000
Model:	OLS	Adj. R-squared (uncentered):	1.000
Method:	Least Squares	F-statistic:	1.063e+05
Date:	Mon, 04 Sep 2023	<pre>Prob (F-statistic):</pre>	0.00
Time:	23:18:25	Log-Likelihood:	1528.0
No. Observations:	569	AIC:	-3018.
Df Residuals:	550	BIC:	-2935.

Df Model: 19

Covariance Type:	nonrobu	st				
=======================================	coef	std err	t	P> t	[0.025	0.975]
texture_mean	-0.0011	0.001	-1.408	0.160	-0.003	0.000
perimeter_mean	0.9366	0.014	64.650	0.000	0.908	0.965
area_mean	0.0802	0.012	6.880	0.000	0.057	0.103
smoothness_mean	0.0092	0.002	5.144	0.000	0.006	0.013
compactness_mean	-0.0560	0.004	-14.924	0.000	-0.063	-0.049
concavity_mean	-0.0343	0.003	-11.829	0.000	-0.040	-0.029
symmetry_mean	0.0028	0.001	2.082	0.038	0.000	0.005
<pre>fractal_dimension_mean</pre>	0.0027	0.002	1.219	0.223	-0.002	0.007
radius_se	0.0092	0.005	1.761	0.079	-0.001	0.019
perimeter_se	-0.0199	0.005	-3.861	0.000	-0.030	-0.010
smoothness_se	0.0027	0.001	2.431	0.015	0.001	0.005
concavity_se	0.0138	0.002	8.525	0.000	0.011	0.017
<pre>fractal_dimension_se</pre>	-0.0030	0.002	-1.912	0.056	-0.006	8.08e-05
radius_worst	0.2180	0.015	14.204	0.000	0.188	0.248
perimeter_worst	-0.0961	0.013	-7.535	0.000	-0.121	-0.071
area_worst	-0.0858	0.009	-9.086	0.000	-0.104	-0.067
smoothness_worst	-0.0080	0.002	-4.310	0.000	-0.012	-0.004

compactness_worst symmetry_worst 	0.0125 -0.0022 	0.002 5.654 0.001 -1.584	0.000 0.114	0.008 -0.005	0.017 0.001
Omnibus: Prob(Omnibus):	72.913 0.000	Durbin-Watson: Jarque-Bera (JB):		1.922 628.194	
Skew: Kurtosis:	-0.107 8.143	Prob(JB): Cond. No.		3.88e-137 95.0	

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

--- La variable eliminada fue: fractal_dimension_mean

Dep. Variable: Model: Method: Date: Time: No. Observations: Df Residuals: Df Model:	radius_ Least Squ Mon, 04 Sep 23:1 nonro	OLS ares 2023 8:25 569 551 18	•	tistic):	ed):	1.000 1.000 1.121e+05 0.00 1527.2 -3018. -2940.
Covariance Type:	 coef	====== std e	======== rr	t P> t	======== [0.025	 0.975]
texture_mean	 -0.0012	 0.0				 0.000
perimeter_mean	0.9307	0.0				0.958
area mean	0.0837	0.0				0.106
smoothness_mean	0.0095	0.0				0.013
compactness mean	-0.0535	0.0	03 -16.97	74 0.000	-0.060	-0.047
concavity_mean	-0.0339	0.0	93 - 11. 76	0.000	-0.040	-0.028
symmetry mean	0.0027	0.0	2.0 3	33 0.043	9.12e-05	0.005
radius_se	0.0079	0.0	1. 54	0.123	-0.002	0.018
perimeter_se	-0.0189	0.0	os -3.76	0.000	-0.029	-0.009
smoothness_se	0.0026	0.0	2.27	77 0.023	0.000	0.005
concavity_se	0.0135	0.0	8.43	9 0.000	0.010	0.017

<pre>fractal_dimension_se</pre>	-0.0019	0.001	-1.485	0.138	-0.004	0.001
radius_worst	0.2211	0.015	14.609	0.000	0.191	0.251
perimeter_worst	-0.0981	0.013	-7.743	0.000	-0.123	-0.073
area_worst	-0.0865	0.009	-9.181	0.000	-0.105	-0.068
smoothness_worst	-0.0078	0.002	-4.222	0.000	-0.011	-0.004
compactness_worst	0.0120	0.002	5.521	0.000	0.008	0.016
symmetry_worst	-0.0021	0.001	-1.515	0.130	-0.005	0.001
Omnibus:	 !.71	======== 500 Durb:	======= in-Watson:	=======	1.914	
Prob(Omnibus):	0.6	000 Jarq	ue-Bera (JB)	:	603.404	
Skew:	-0.6	990 Prob	(JB):		9.39e-132	
Kurtosis:	8.6	042 Cond	. No.		91.3	
=======================================	=========			========	========	

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

--- La variable eliminada fue: texture_mean

	'	ols ke	gress	TOU KESUICS			
Dep. Variable:	radius_	 mean	 R-sq	uared (unce	 ntered):		1.000
Model:	_	OLS	Adj.	R-squared	(uncentered):	1.000
Method:	Least Squ	ares	F-st	atistic:			1.184e+05
Date:	Mon, 04 Sep	2023	Prob	(F-statist	ic):		0.00
Time:	23:1	8:25	Log-	Likelihood:	·		1526.1
No. Observations:		569	AIC:				-3018.
Df Residuals:		552	BIC:				-2944.
Df Model:		17					
Covariance Type:	nonro	bust					
=======================================	coef	===== std	==== err 	t	P> t	[0.025	0.975]
perimeter_mean	0.9307	0.	 014	68.019	0.000	0.904	0.958
area_mean	0.0842	0.	011	7.442	0.000	0.062	0.106
smoothness_mean	0.0099	0.	002	5.702	0.000	0.007	0.013
compactness_mean	-0.0533	0.	003	-16.909	0.000	-0.060	-0.047
concavity_mean	-0.0344	0.	003	-11.952	0.000	-0.040	-0.029
symmetry_mean	0.0026	0.	001	1.978	0.048	1.85e-05	0.005

radius_se	0.0081	0.005	1.589	0.113	-0.002	0.018
perimeter_se	-0.0192	0.005	-3.781	0.000	-0.029	-0.009
smoothness_se	0.0023	0.001	2.084	0.038	0.000	0.004
concavity_se	0.0136	0.002	8.550	0.000	0.010	0.017
<pre>fractal_dimension_se</pre>	-0.0019	0.001	-1.449	0.148	-0.004	0.001
radius_worst	0.2194	0.015	14.523	0.000	0.190	0.249
perimeter_worst	-0.0972	0.013	-7.675	0.000	-0.122	-0.072
area_worst	-0.0863	0.009	-9.145	0.000	-0.105	-0.068
smoothness_worst	-0.0079	0.002	-4.246	0.000	-0.012	-0.004
compactness_worst	0.0116	0.002	5.382	0.000	0.007	0.016
symmetry_worst	-0.0020	0.001	-1.437	0.151	-0.005	0.001
=======================================	========	=======	========	=======	========	
Omnibus:	73.	605 Durbi	in-Watson:		1.916	
Prob(Omnibus):	0.	000 Jarqı	ue-Bera (JB)	:	651.590	
Skew:	-0.	090 Prob	(JB):		3.23e-142	
Kurtosis:	8.	239 Cond	. No.		90.6	
=======================================	========	========	========	========	========	

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

--- La variable eliminada fue: symmetry_worst

=======================================	============	======	========	:=======	=======	:=======
Dep. Variable:	radius mea	n R-s	quared (unce	entered):		1.000
Model:	_ OL		. R-squared	(uncentered)	:	1.000
Method:	Least Square	s F-s	tatistic:			1.256e+05
Date:	Mon, 04 Sep 202	3 Pro	b (F-statist	ic):		0.00
Time:	23:18:2	5 Log	-Likelihood:			1525.0
No. Observations:	56	9 AIC	•			-3018.
Df Residuals:	55	3 BIC	. •			-2949.
Df Model:	1	6				
Covariance Type:	nonrobus	t				
=======================================	=======================================	======	========	:========	=======	========
	coef s	td err	t	P> t	[0.025	0.975]
perimeter_mean	0.9347	0.013	69.692	0.000	0.908	0.961
area_mean	0.0814	0.011	7.297	0.000	0.059	0.103

smoothness_mean	0.0102	0.002	5.853	0.000	0.007	0.014
compactness_mean	-0.0531	0.003	-16.847	0.000	-0.059	-0.047
concavity_mean	-0.0343	0.003	-11.926	0.000	-0.040	-0.029
symmetry_mean	0.0013	0.001	1.361	0.174	-0.001	0.003
radius_se	0.0087	0.005	1.701	0.090	-0.001	0.019
perimeter_se	-0.0195	0.005	-3.824	0.000	-0.029	-0.009
smoothness_se	0.0025	0.001	2.305	0.022	0.000	0.005
concavity_se	0.0135	0.002	8.502	0.000	0.010	0.017
<pre>fractal_dimension_se</pre>	-0.0017	0.001	-1.308	0.191	-0.004	0.001
radius_worst	0.2154	0.015	14.494	0.000	0.186	0.245
perimeter_worst	-0.0970	0.013	-7.654	0.000	-0.122	-0.072
area_worst	-0.0836	0.009	-9.029	0.000	-0.102	-0.065
smoothness_worst	-0.0082	0.002	-4.466	0.000	-0.012	-0.005
compactness_worst	0.0108	0.002	5.182	0.000	0.007	0.015
=======================================	========	=======	========	========	========	
Omnibus:	77.	721 Durb	in-Watson:		1.911	
Prob(Omnibus):	0.	000 Jarq	ue-Bera (JB)	:	718.228	
Skew:	-0.	153 Prob	(JB):		1.09e-156	
Kurtosis:	8.	496 Cond	. No.		87.6	

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

--- La variable eliminada fue: fractal_dimension_se

coef

std err

OLS Regression Results

Dep. Variable:	radius_mean	R-squared (uncentered):	1.000				
Model:	OLS	Adj. R-squared (uncentered):	1.000				
Method:	Least Squares	F-statistic:	1.338e+05				
Date:	Mon, 04 Sep 2023	<pre>Prob (F-statistic):</pre>	0.00				
Time:	23:18:25	Log-Likelihood:	1524.2				
No. Observations:	569	AIC:	-3018.				
Df Residuals:	554	BIC:	-2953.				
Df Model:	15						
Covariance Type:	nonrobust						
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P>|t|

[0.025

0.975]

perimeter_mean	0.9356	0.013	69.794	0.000	0.909	0.962
area_mean	0.0808	0.011	7.243	0.000	0.059	0.103
smoothness_mean	0.0102	0.002	5.859	0.000	0.007	0.014
compactness_mean	-0.0544	0.003	-18.193	0.000	-0.060	-0.049
concavity_mean	-0.0336	0.003	-11.898	0.000	-0.039	-0.028
symmetry_mean	0.0014	0.001	1.456	0.146	-0.000	0.003
radius_se	0.0078	0.005	1.539	0.124	-0.002	0.018
perimeter_se	-0.0187	0.005	-3.691	0.000	-0.029	-0.009
smoothness_se	0.0022	0.001	2.057	0.040	9.93e-05	0.004
concavity_se	0.0125	0.001	9.088	0.000	0.010	0.015
radius_worst	0.2152	0.015	14.478	0.000	0.186	0.244
perimeter_worst	-0.0967	0.013	-7.623	0.000	-0.122	-0.072
area_worst	-0.0837	0.009	-9.031	0.000	-0.102	-0.065
smoothness_worst	-0.0079	0.002	-4.324	0.000	-0.012	-0.004
compactness_worst	0.0107	0.002	5.130	0.000	0.007	0.015
Omnibus:		74.569	====== Durbin-Wats	======= on:	: 1	==== .911
Prob(Omnibus):		0.000	Jarque-Bera			.014
Skew:		-0.109	Prob(JB):	().	2.38e	
Kurtosis:		8.296	Cond. No.			87 . 1
=======================================	:=======	:======	========	========	========	====

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

--- La variable eliminada fue: symmetry_mean

===========	=======================================		=========
Dep. Variable:	radius_mean	R-squared (uncentered):	1.000
Model:	OLS	Adj. R-squared (uncentered):	1.000
Method:	Least Squares	F-statistic:	1.431e+05
Date:	Mon, 04 Sep 2023	<pre>Prob (F-statistic):</pre>	0.00
Time:	23:18:25	Log-Likelihood:	1523.1
No. Observations:	569	AIC:	-3018.
Df Residuals:	555	BIC:	-2957.
Df Model:	14		
Covariance Type:	nonrobust		

==========	========	=======	========	========		=======	
	coef	std err	t	P> t	[0.025	0.975]	
perimeter_mean	0.9342	0.013	69.796	0.000	0.908	0.960	
area_mean	0.0813	0.011	7.290	0.000	0.059	0.103	
smoothness_mean	0.0106	0.002	6.144	0.000	0.007	0.014	
compactness_mean	-0.0536	0.003	-18.211	0.000	-0.059	-0.048	
concavity_mean	-0.0332	0.003	-11.809	0.000	-0.039	-0.028	
radius_se	0.0085	0.005	1.676	0.094	-0.001	0.018	
perimeter_se	-0.0192	0.005	-3.800	0.000	-0.029	-0.009	
smoothness_se	0.0022	0.001	2.027	0.043	6.73e-05	0.004	
concavity_se	0.0124	0.001	9.009	0.000	0.010	0.015	
radius_worst	0.2157	0.015	14.494	0.000	0.186	0.245	
perimeter_worst	-0.0959	0.013	-7.560	0.000	-0.121	-0.071	
area_worst	-0.0845	0.009	-9.133	0.000	-0.103	-0.066	
smoothness_worst	-0.0081	0.002	-4.435	0.000	-0.012	-0.005	
compactness_worst	0.0106	0.002	5.075	0.000	0.006	0.015	
Omnibus:	========	72.226	Durbin-Watson: 1.910		=== 916		
<pre>Prob(Omnibus):</pre>		0.000	Jarque-Bera (JB):		622.372		
Skew:		-0.082	• • • • • • • • • • • • • • • • • • • •		7.14e-		
Kurtosis:		8.121	Cond. No. 86.0				

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

--- La variable eliminada fue: radius_se

Dep. Variable:	radius_mean	R-squared (uncentered):	1.000
Model:	OLS	Adj. R-squared (uncentered):	1.000
Method:	Least Squares	F-statistic:	1.536e+05
Date:	Mon, 04 Sep 2023	<pre>Prob (F-statistic):</pre>	0.00
Time:	23:18:25	Log-Likelihood:	1521.6
No. Observations:	569	AIC:	-3017.
Df Residuals:	556	BIC:	-2961.
Df Model:	13		

Covariance Type:	nonrobust					
	coef	std err	t	P> t	[0.025	0.975]
perimeter_mean	0.9314	0.013	70.006	0.000	0.905	0.958
area_mean	0.0821	0.011	7.347	0.000	0.060	0.104
smoothness_mean	0.0110	0.002	6.459	0.000	0.008	0.014
compactness_mean	-0.0538	0.003	-18.234	0.000	-0.060	-0.048
concavity_mean	-0.0324	0.003	-11.681	0.000	-0.038	-0.027
perimeter_se	-0.0112	0.002	-7.032	0.000	-0.014	-0.008
smoothness_se	0.0025	0.001	2.362	0.019	0.000	0.005
concavity_se	0.0123	0.001	8.932	0.000	0.010	0.015
radius_worst	0.2313	0.012	19.972	0.000	0.209	0.254
perimeter_worst	-0.1092	0.010	-11.032	0.000	-0.129	-0.090
area_worst	-0.0850	0.009	-9.174	0.000	-0.103	-0.067
smoothness_worst	-0.0085	0.002	-4.698	0.000	-0.012	-0.005
compactness_worst	0.0106	0.002	5.072	0.000	0.006	0.015
Omnibus:	=======	75.002	 Durbin-Watson:		1.910	
Prob(Omnibus):		0.000	Jarque-Bera (JB):		675.381	
Skew:		-0.111	Prob(JB):		2.20e-147	
Kurtosis:		8.333	Cond. No.		79.	8

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

6.-Comparación entre datos reales y predicción. Análisis de los resultados.

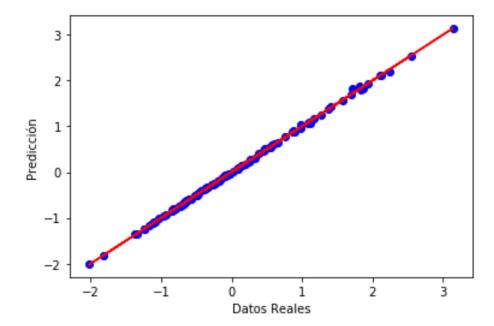
El modelo de regresión múltiple muestra un R-squared perfecto, lo que indica que el modelo se ajusta perfectamente a los datos, asi mismo se puede determinar que los coeficientes de las variables independientes son todos significativos.

A continuación este análisis encaja en veracidad, al ver como la gráfica de valores de predicción y valores reales se alinea con la línea roja porque esto significa que el modelo de regresión está haciendo predicciones muy precisas y que los valores predichos son prácticamente idénticos a los valores reales. También se realiza el grafico e histograma de los residuos. Y un qqplot el cual sus puntos se ven muy cercanos a la línea.

```
In [12]: y_aprox=modelo2.params[0]*prueba['perimeter_mean']+modelo2.params[1]*prueba['area_mean']+modelo
2.params[2]*prueba['smoothness_mean']+modelo2.params[3]*prueba['compactness_mean']+modelo2.param
s[4]*prueba['concavity_mean']+modelo2.params[5]*prueba['perimeter_se']+modelo2.params[6]*prueba
['smoothness_se']+modelo2.params[7]*prueba['concavity_se']+modelo2.params[8]*prueba['radius_wors
t']+modelo2.params[9]*prueba['perimeter_worst']+modelo2.params[10]*prueba['area_worst']+modelo2.
params[11]*prueba['smoothness_worst']+modelo2.params[12]*prueba['compactness_worst']
tabla=pd.DataFrame({'Real': prueba['radius_mean'], 'Prediccion': y_aprox, 'Errores': prueba['radius_mean']-y_aprox})
```

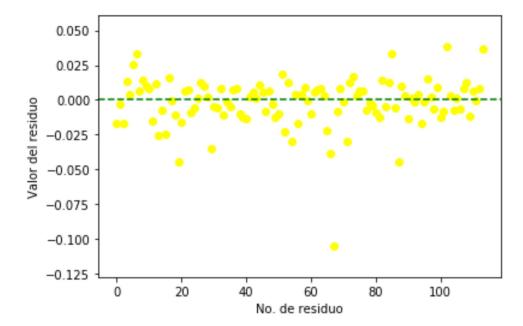
```
In [13]: plt.scatter(prueba['radius_mean'], y_aprox, color='blue')
  plt.plot(prueba['radius_mean'], prueba['radius_mean'], color='red')
  plt.xlabel("Datos Reales")
  plt.ylabel("Predicción")
```

Out[13]: Text(0, 0.5, 'Predicción')



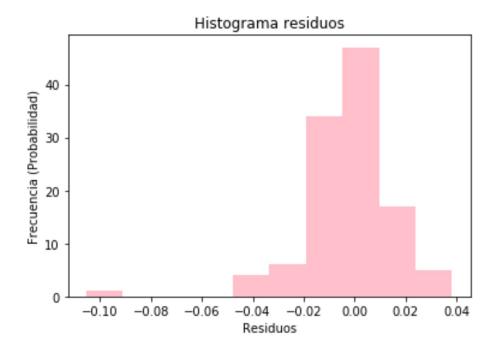
```
In [14]: plt.scatter(range(tabla.shape[0]),tabla['Errores'], color='yellow')
    plt.axhline(y=0, linestyle='--', color='green')
    plt.xlabel("No. de residuo")
    plt.ylabel("Valor del residuo")
```

Out[14]: Text(0, 0.5, 'Valor del residuo')



```
In [15]: plt.hist(x=tabla['Errores'], color='pink')
    plt.title('Histograma residuos')
    plt.xlabel("Residuos")
    plt.ylabel("Frecuencia (Probabilidad)")
```

Out[15]: Text(0, 0.5, 'Frecuencia (Probabilidad)')



In [16]: QQ = sm.qqplot(tabla['Errores'], stats.norm, line='s')

