

Regresiones lineales

El propósito de este script es determinar si un modelo lineal es adecuado para construir una función analítica entre el salario y las medidas del esfuerzo. Esto dependerá de la significancia estadística de las variables, así como el número que haya de estas en cada modelo.

```
In [1]: from sklearn.preprocessing import StandardScaler
import statsmodels.api as sm
import pandas as pd
import numpy as np
import math
import os
import warnings
print('Modulos importados')
```

Modulos importados

```
In [2]: # Configuraciones
warnings.filterwarnings('ignore')
# Reduzcamos el número de línea a leer
pd.options.display.max_rows = 5
```

```
In [3]: # Años de análisis
period = 11
csv = '.csv'
# Directorios
pitcher_path = '../Data/New_Data/Free_Agent/Pitchers/complete_free_agents_pitchers_'
hitter_path = '../Data/New_Data/Free_Agent/Hitters/complete_free_agents_hitters_'
# Originales:
df_pitcher = [None]*period
df_hitter = [None]*period
```

```
In [4]: # Función para normalizar datos
def normalizar(df):
    df_normalizado = df.copy()
    for etiqueta in df.columns:
        maximo = df[etiqueta].max()
        minimo = df[etiqueta].min()
        df_normalizado[etiqueta] = (df[etiqueta] - minimo) / (maximo - minimo)

    return df_normalizado
```

```
In [5]: for i in range(0, period):
        df_pitcher[i] = pd.read_csv(pitcher_path + str(2011 + i) + csv)
        df_hitter[i] = pd.read_csv(hitter_path + str(2011 + i) + csv)
```

Veamos el tipo de datos contenidos en las bases de datos

```
In [6]: df_pitcher[2].head()
```

```
Out[6]:
```

	Jugador	Valor_contrato	Valor_promedio_contrato	Posicion	Juegos	Juegos_iniciados	Inning_pitched	Bateos	Carreras	Carreras_ganadas
0	Zack Greinke	147000000	24500000	SP	29	28	177.7	152	54	
1	J.P. Howell	2850000	2850000	RP	67	0	62.0	42	14	
2	Brian Wilson	1000000	1000000	RP	18	0	13.7	8	1	
3	Peter Moylan	0	0	RP	14	0	15.3	23	11	
4	Anibal Sanchez	80000000	16000000	RP	29	29	182.0	156	56	

5 rows × 22 columns

```
In [7]: df_hitter[8].head()
```

Out[7]:

	Jugador	Valor_contrato	Valor_promedio_contrato	Posicion	Juegos	Porcetnaje_juegos	At-bats	Bateos	Home-runs	RBI	Porcentaje_bateo
0	Bryce Harper	330000000	25384615	RF	157	0.969	573	149	35	114	0.260
1	Andrew McCutchen	50000000	16666667	RF	59	0.364	219	56	10	29	0.256
2	Blake Parker	0	0	RP	60	0.370	1	0	0	0	0.000
3	Manny Machado	300000000	30000000	3B	156	0.963	587	150	32	85	0.256
4	Garrett Richards	15500000	7750000	SP	3	0.019	2	0	0	0	0.000

Debido a la enorme diferencias en la magnitud entre el promedio de victorias y los sueldos, se tendrá que estadandarizar los sueldos. A pesar de que intuitivamente no tenemos que usar la columna del valor de contrato, también estandarizémosla.

In [9]:

```
# Originales:
df_pitcher_copy = [None]*period
df_hitter_copy = [None]*period
```

```

In [12]:
for i in range(1,period):
    df_pitcher_copy[i] = df_pitcher[i].copy()
    df_hitter_copy[i] = df_hitter[i].copy()

    maximo_p_1 = df_pitcher_copy[i]['Sueldo'].max()
    minimo_p_1 = df_pitcher_copy[i]['Sueldo'].min()

    maximo_h_1 = df_hitter_copy[i]['Sueldo'].max()
    minimo_h_1 = df_hitter_copy[i]['Sueldo'].min()

    df_pitcher_copy[i]['Sueldo_estandarizado'] = (df_pitcher_copy[i]['Sueldo'] - minimo_p_1)/(maximo_p_1 - minimo_p_1)
    df_hitter_copy[i]['Sueldo_estandarizado'] = (df_hitter_copy[i]['Sueldo'] - minimo_h_1)/(maximo_h_1 - minimo_h_1)

    maximo_p_2 = df_pitcher_copy[i]['Valor_contrato'].max()
    minimo_p_2 = df_pitcher_copy[i]['Valor_contrato'].min()

    maximo_h_2 = df_hitter_copy[i]['Valor_contrato'].max()
    minimo_h_2 = df_hitter_copy[i]['Valor_contrato'].min()

    df_pitcher_copy[i]['Valor_contrato_estandarizado'] = (df_pitcher_copy[i]['Valor_contrato'] - minimo_p_2)/(maximo_p_2 - minimo_p_2)
    df_hitter_copy[i]['Valor_contrato_estandarizado'] = (df_hitter_copy[i]['Valor_contrato'] - minimo_h_2)/(maximo_h_2 - minimo_h_2)

```

```

In [13]: df_hitter_copy[3].head()

```

```

Out[13]:

```

	Jugador	Valor_contrato	Valor_promedio_contrato	Posicion	Juegos	Porcetnaje_juegos	At-bats	Bateos	Home-runs	RBI	Porcentaje_bateo
0	Robinson Cano	240000000	24000000	2B	157	0.969	595	187	14	82	0.314
1	Kendrys Morales	7540983	7540983	DH	98	0.605	367	80	8	42	0.218
2	Chris Young	7250000	7250000	RF	111	0.685	325	72	11	38	0.222
3	Chris Young	7250000	7250000	RP	30	0.185	1	0	0	0	0.000
4	Corey Hart	6000000	6000000	RF	68	0.420	232	47	6	21	0.203

Ahora, tanto la variable dependiente como independiente están en el intervalo $[0,1]$

```
In [14]: df_pitcher_copy[2].head()
```

```
Out[14]:
```

	Jugador	Valor_contrato	Valor_promedio_contrato	Posicion	Juegos	Juegos_iniciados	Inning_pitched	Bateos	Carreras	Carreras_ganadas
0	Zack Greinke	147000000	24500000	SP	29	28	177.7	152	54	
1	J.P. Howell	2850000	2850000	RP	67	0	62.0	42	14	
2	Brian Wilson	1000000	1000000	RP	18	0	13.7	8	1	
3	Peter Moylan	0	0	RP	14	0	15.3	23	11	
4	Anibal Sanchez	80000000	16000000	RP	29	29	182.0	156	56	

5 rows × 11 columns

I. Victorias en función del salario

$$\Delta_t = a + b\omega_t + e_t$$

donde Δ_t son las victorias por temporada. Se aplicará el modelo tanto para agentes libres que sean *hitters* o *pitchers*. Para cada año de análisis, consideraremos el promedio de victorias como medida del esfuerzo y el total del salario, para después hacer con el valor del contrato.

Creemos una lista para guardar los modelos y proceder a obtenerlos

```
In [15]: # Modelos
hitter_model_I_a = [None]*period
pitcher_model_I_a = [None]*period
```

```
In [16]: for i in range(1,period):
# Hitters:
y_hitter = df_hitter_copy[i]['Promedio_victorias'].tolist()
x_hitter = df_hitter_copy[i]['Sueldo_estandarizado'].tolist()

x_hitter = sm.add_constant(x_hitter)

hitter_model_I_a[i] = sm.OLS(y_hitter, x_hitter).fit()

# Pitchers:
y_pitcher = df_pitcher_copy[i]['Promedio_victorias'].tolist()
x_pitcher = df_pitcher_copy[i]['Sueldo_estandarizado'].tolist()

x_pitcher = sm.add_constant(x_pitcher)

pitcher_model_I_a[i] = sm.OLS(y_pitcher, x_pitcher).fit()
```

Observemos los resúmenes de las regresiones

Hitters

```
In [17]: for i in range(1,period):
print(hitter_model_I_a[i].summary())
```

```

OLS Regression Results
=====
Dep. Variable:          y      R-squared:          0.000
Model:                  OLS    Adj. R-squared:       -0.021
Method:                 Least Squares  F-statistic:        0.001698
Date:                   Wed, 29 Jun 2022  Prob (F-statistic):    0.967
Time:                   09:46:46   Log-Likelihood:      65.615
No. Observations:       49      AIC:                -127.2
Df Residuals:           47      BIC:                -123.4
Df Model:                1
Covariance Type:        nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.5191	0.013	41.275	0.000	0.494	0.544
x1	-0.0019	0.047	-0.041	0.967	-0.096	0.092

```
=====
```

Omnibus:	4.196	Durbin-Watson:	1.365
Prob(Omnibus):	0.123	Jarque-Bera (JB):	3.963
Skew:	-0.641	Prob(JB):	0.138
Kurtosis:	2.457	Cond. No.	5.25

Notes:

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

OLS Regression Results

Dep. Variable:	y	R-squared:	0.040
Model:	OLS	Adj. R-squared:	0.028
Method:	Least Squares	F-statistic:	3.199
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.0777
Time:	09:46:46	Log-Likelihood:	98.048
No. Observations:	78	AIC:	-192.1
Df Residuals:	76	BIC:	-187.4
Df Model:	1		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
const	0.4972	0.011	46.251	0.000	0.476	0.519
x1	0.0576	0.032	1.789	0.078	-0.007	0.122

Omnibus:	9.063	Durbin-Watson:	0.624
Prob(Omnibus):	0.011	Jarque-Bera (JB):	5.432
Skew:	-0.475	Prob(JB):	0.0661
Kurtosis:	2.123	Cond. No.	4.30

Notes:

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

OLS Regression Results

Dep. Variable:	y	R-squared:	0.000
Model:	OLS	Adj. R-squared:	-0.011
Method:	Least Squares	F-statistic:	0.04023
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.841
Time:	09:46:46	Log-Likelihood:	134.37
No. Observations:	92	AIC:	-264.7
Df Residuals:	90	BIC:	-259.7
Df Model:	1		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
-----	-----	-----	-----	-----	-----	-----
const	0.5042	0.008	62.062	0.000	0.488	0.520
x1	-0.0053	0.026	-0.201	0.841	-0.058	0.047
=====	=====	=====	=====	=====	=====	=====
Omnibus:		11.983	Durbin-Watson:			0.727
Prob(Omnibus):		0.002	Jarque-Bera (JB):			3.771
Skew:		-0.026	Prob(JB):			0.152
Kurtosis:		2.010	Cond. No.			4.66
=====	=====	=====	=====	=====	=====	=====

Notes:

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

OLS Regression Results

Dep. Variable:	y	R-squared:	0.001
Model:	OLS	Adj. R-squared:	-0.012
Method:	Least Squares	F-statistic:	0.09970
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.753
Time:	09:46:46	Log-Likelihood:	101.08
No. Observations:	79	AIC:	-198.2
Df Residuals:	77	BIC:	-193.4
Df Model:	1		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
-----	-----	-----	-----	-----	-----	-----
const	0.5159	0.010	51.559	0.000	0.496	0.536
x1	0.0092	0.029	0.316	0.753	-0.049	0.067
=====	=====	=====	=====	=====	=====	=====
Omnibus:		13.836	Durbin-Watson:			0.688
Prob(Omnibus):		0.001	Jarque-Bera (JB):			4.995
Skew:		-0.325	Prob(JB):			0.0823
Kurtosis:		1.954	Cond. No.			4.00
=====	=====	=====	=====	=====	=====	=====

Notes:

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

OLS Regression Results

Dep. Variable:	y	R-squared:	0.063
Model:	OLS	Adj. R-squared:	0.054
Method:	Least Squares	F-statistic:	7.132
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.00877
Time:	09:46:46	Log-Likelihood:	157.21


```

No. Observations:      108   AIC:      -310.4
Df Residuals:         106   BIC:      -305.1
Df Model:              1
Covariance Type:      nonrobust

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.5080	0.007	73.304	0.000	0.494	0.522
x1	0.0792	0.030	2.671	0.009	0.020	0.138
Omnibus:	0.752		Durbin-Watson:		0.673	
Prob(Omnibus):	0.687		Jarque-Bera (JB):		0.849	
Skew:	-0.185		Prob(JB):		0.654	
Kurtosis:	2.772		Cond. No.		5.52	

Notes:

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

OLS Regression Results

```

Dep. Variable:      y   R-squared:      0.013
Model:              OLS   Adj. R-squared:  -0.003
Method:             Least Squares   F-statistic:  0.8014
Date:               Wed, 29 Jun 2022   Prob (F-statistic): 0.374
Time:               09:46:46   Log-Likelihood: 82.715
No. Observations:   64   AIC:      -161.4
Df Residuals:       62   BIC:      -157.1
Df Model:           1
Covariance Type:    nonrobust

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.5060	0.013	39.648	0.000	0.481	0.532
x1	0.0346	0.039	0.895	0.374	-0.043	0.112
Omnibus:	8.269		Durbin-Watson:		0.556	
Prob(Omnibus):	0.016		Jarque-Bera (JB):		4.219	
Skew:	0.410		Prob(JB):		0.121	
Kurtosis:	2.046		Cond. No.		4.88	

Notes:

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

OLS Regression Results

```

Dep. Variable:          y      R-squared:          0.002
Model:                OLS     Adj. R-squared:       -0.016
Method:              Least Squares   F-statistic:        0.1261
Date:                Wed, 29 Jun 2022   Prob (F-statistic):   0.724
Time:                09:46:46   Log-Likelihood:      64.263
No. Observations:      58      AIC:                -124.5
Df Residuals:          56      BIC:                -120.4
Df Model:              1
Covariance Type:      nonrobust

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.5284	0.015	35.781	0.000	0.499	0.558
x1	-0.0174	0.049	-0.355	0.724	-0.115	0.081
Omnibus:	7.720		Durbin-Watson:	0.882		
Prob(Omnibus):	0.021		Jarque-Bera (JB):	6.879		
Skew:	-0.754		Prob(JB):	0.0321		
Kurtosis:	3.756		Cond. No.	4.80		

Notes:

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

OLS Regression Results

Dep. Variable:	y	R-squared:	0.026			
Model:	OLS	Adj. R-squared:	0.010			
Method:	Least Squares	F-statistic:	1.657			
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.203			
Time:	09:46:46	Log-Likelihood:	74.775			
No. Observations:	65	AIC:	-145.5			
Df Residuals:	63	BIC:	-141.2			
Df Model:	1					
Covariance Type:	nonrobust					
=====						
	coef	std err	t	P> t	[0.025	0.975]

const	0.5329	0.014	38.382	0.000	0.505	0.561
x1	0.0509	0.040	1.287	0.203	-0.028	0.130
=====						
Omnibus:	7.580	Durbin-Watson:	0.852			
Prob(Omnibus):	0.023	Jarque-Bera (JB):	6.898			
Skew:	-0.758	Prob(JB):	0.0318			
Kurtosis:	3.499	Cond. No.	4.38			

Notes:

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

OLS Regression Results

Dep. Variable:	y	R-squared:	0.076
Model:	OLS	Adj. R-squared:	0.056
Method:	Least Squares	F-statistic:	3.767
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.0584
Time:	09:46:46	Log-Likelihood:	56.818
No. Observations:	48	AIC:	-109.6
Df Residuals:	46	BIC:	-105.9
Df Model:	1		
Covariance Type:	nonrobust		
=====			
	coef	std err	t
			P> t
			[0.025
			0.975]

const	0.4797	0.015	31.572
			0.000
x1	0.0844	0.043	1.941
			0.058

Omnibus:	5.392	Durbin-Watson:	1.209
Prob(Omnibus):	0.067	Jarque-Bera (JB):	4.553
Skew:	0.744	Prob(JB):	0.103
Kurtosis:	3.251	Cond. No.	4.23
=====			

Notes:

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

OLS Regression Results

=====						
Dep. Variable:	y	R-squared:	0.058			
Model:	OLS	Adj. R-squared:	0.046			
Method:	Least Squares	F-statistic:	4.531			
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.0367			
Time:	09:46:46	Log-Likelihood:	92.286			
No. Observations:	75	AIC:	-180.6			
Df Residuals:	73	BIC:	-175.9			
Df Model:	1					
Covariance Type:	nonrobust					
=====						
	coef	std err	t	P> t	[0.025	0.975]

const	0.5193	0.011	46.328	0.000	0.497	0.542
x1	0.0914	0.043	2.129	0.037	0.006	0.177
=====						

Omnibus:	0.369	Durbin-Watson:	0.742
Prob(Omnibus):	0.832	Jarque-Bera (JB):	0.535
Skew:	0.029	Prob(JB):	0.765
Kurtosis:	2.590	Cond. No.	5.36

Notes:

Los resultados muestran que, en general, hay poca significancia estadística para explicar las victorias con el salario de los agentes libres. Así mismo, tampoco tiene poder predictivo puesto que la R^2 es baja. Faltará corroborar esto con validación cruzada, sin embargo, el hecho de la baja significancia estadística no cambiará.

Pitchers

```
In [44]: for i in range(1,period):
          print(pitcher_model_I_a[i].summary())
```

```

OLS Regression Results
=====
Dep. Variable:          y      R-squared:          0.007
Model:                OLS      Adj. R-squared:       -0.051
Method:             Least Squares  F-statistic:        0.1269
Date:                Wed, 29 Jun 2022  Prob (F-statistic):    0.726
Time:                  01:05:24  Log-Likelihood:      25.191
No. Observations:      19      AIC:                -46.38
Df Residuals:          17      BIC:                -44.49
Df Model:               1
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.5194	0.027	19.110	0.000	0.462	0.577
x1	0.0168	0.047	0.356	0.726	-0.083	0.116

```

=====
Omnibus:                2.770    Durbin-Watson:          1.289
Prob(Omnibus):           0.250    Jarque-Bera (JB):        1.560
Skew:                   -0.422    Prob(JB):                0.459
Kurtosis:                1.879    Cond. No.:               3.77
=====

```

Notes:

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

OLS Regression Results

```

=====
Dep. Variable:          y    R-squared:          0.000
Model:                OLS   Adj. R-squared:       -0.020
Method:              Least Squares   F-statistic:        0.003177
Date:                Wed, 29 Jun 2022   Prob (F-statistic):    0.955
Time:                01:05:24   Log-Likelihood:       63.550
No. Observations:      51   AIC:                 -123.1
Df Residuals:          49   BIC:                 -119.2
Df Model:              1
Covariance Type:      nonrobust
=====

```

```

=====
              coef      std err          t      P>|t|      [0.025      0.975]
-----
const         0.4999      0.014     35.082      0.000      0.471      0.529
x1            0.0027      0.047      0.056      0.955     -0.093      0.098
=====

```

```

=====
Omnibus:                3.936   Durbin-Watson:          0.591
Prob(Omnibus):           0.140   Jarque-Bera (JB):        3.242
Skew:                   -0.506   Prob(JB):                0.198
Kurtosis:               2.291   Cond. No.                5.00
=====

```

Notes:

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

OLS Regression Results

```

=====
Dep. Variable:          y    R-squared:          0.011
Model:                OLS   Adj. R-squared:       -0.004
Method:              Least Squares   F-statistic:        0.7258
Date:                Wed, 29 Jun 2022   Prob (F-statistic):    0.397
Time:                01:05:24   Log-Likelihood:       95.650
No. Observations:      68   AIC:                 -187.3
Df Residuals:          66   BIC:                 -182.9
Df Model:              1
Covariance Type:      nonrobust
=====

```

```

=====
              coef      std err          t      P>|t|      [0.025      0.975]
-----
const         0.4842      0.011     45.026      0.000      0.463      0.506
x1            0.0246      0.029      0.852      0.397     -0.033      0.082
=====

```

```

=====
Omnibus:                12.905   Durbin-Watson:          1.153
Prob(Omnibus):           0.002   Jarque-Bera (JB):        3.681
Skew:                   0.141   Prob(JB):                0.159
Kurtosis:               1.896   Cond. No.                4.28
=====

```

Notes:

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

OLS Regression Results

```

=====
Dep. Variable:          y      R-squared:          0.047
Model:                  OLS    Adj. R-squared:       0.034
Method:                 Least Squares  F-statistic:      3.483
Date:                  Wed, 29 Jun 2022  Prob (F-statistic): 0.0662
Time:                  01:05:24  Log-Likelihood:   93.291
No. Observations:      72      AIC:             -182.6
Df Residuals:          70      BIC:             -178.0
Df Model:               1
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.4992	0.011	47.178	0.000	0.478	0.520
x1	0.0696	0.037	1.866	0.066	-0.005	0.144

```

=====
Omnibus:                9.701    Durbin-Watson:      0.759
Prob(Omnibus):          0.008    Jarque-Bera (JB):    3.587
Skew:                   -0.217    Prob(JB):            0.166
Kurtosis:               1.996    Cond. No.            4.88
=====

```

Notes:

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

OLS Regression Results

```

=====
Dep. Variable:          y      R-squared:          0.043
Model:                  OLS    Adj. R-squared:       0.030
Method:                 Least Squares  F-statistic:      3.259
Date:                  Wed, 29 Jun 2022  Prob (F-statistic): 0.0751
Time:                  01:05:24  Log-Likelihood:   105.71
No. Observations:      75      AIC:             -207.4
Df Residuals:          73      BIC:             -202.8
Df Model:               1
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.4987	0.009	53.934	0.000	0.480	0.517
x1	0.0690	0.038	1.805	0.075	-0.007	0.145

Omnibus:	3.307	Durbin-Watson:	0.716
Prob(Omnibus):	0.191	Jarque-Bera (JB):	1.891
Skew:	-0.101	Prob(JB):	0.388
Kurtosis:	2.249	Cond. No.	5.67

Notes:

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

OLS Regression Results

Dep. Variable:	y	R-squared:	0.033
Model:	OLS	Adj. R-squared:	0.008
Method:	Least Squares	F-statistic:	1.326
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.257
Time:	01:05:24	Log-Likelihood:	54.701
No. Observations:	41	AIC:	-105.4
Df Residuals:	39	BIC:	-102.0
Df Model:	1		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
const	0.4896	0.018	27.252	0.000	0.453	0.526
x1	0.0515	0.045	1.152	0.257	-0.039	0.142

Omnibus:	2.601	Durbin-Watson:	1.550
Prob(Omnibus):	0.272	Jarque-Bera (JB):	1.575
Skew:	0.207	Prob(JB):	0.455
Kurtosis:	2.134	Cond. No.	4.89

Notes:

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

OLS Regression Results

Dep. Variable:	y	R-squared:	0.000
Model:	OLS	Adj. R-squared:	-0.019
Method:	Least Squares	F-statistic:	0.006386
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.937
Time:	01:05:24	Log-Likelihood:	59.417
No. Observations:	55	AIC:	-114.8
Df Residuals:	53	BIC:	-110.8
Df Model:	1		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
const	0.5220	0.016	31.912	0.000	0.489	0.555
x1	-0.0049	0.061	-0.080	0.937	-0.127	0.117
Omnibus:		8.398	Durbin-Watson:			0.763
Prob(Omnibus):		0.015	Jarque-Bera (JB):			8.164
Skew:		-0.936	Prob(JB):			0.0169
Kurtosis:		3.245	Cond. No.			5.60

Notes:

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

OLS Regression Results

Dep. Variable:	y	R-squared:	0.031
Model:	OLS	Adj. R-squared:	0.010
Method:	Least Squares	F-statistic:	1.472
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.231
Time:	01:05:24	Log-Likelihood:	52.140
No. Observations:	48	AIC:	-100.3
Df Residuals:	46	BIC:	-96.54
Df Model:	1		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
const	0.5162	0.019	27.557	0.000	0.478	0.554
x1	0.0625	0.051	1.213	0.231	-0.041	0.166
Omnibus:		8.856	Durbin-Watson:			1.203
Prob(Omnibus):		0.012	Jarque-Bera (JB):			7.994
Skew:		-0.896	Prob(JB):			0.0184
Kurtosis:		3.886	Cond. No.			4.63

Notes:

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

OLS Regression Results

Dep. Variable:	y	R-squared:	0.001
Model:	OLS	Adj. R-squared:	-0.017
Method:	Least Squares	F-statistic:	0.03318
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.856


```

Time:                01:05:24    Log-Likelihood:        57.720
No. Observations:    58          AIC:                  -111.4
Df Residuals:        56          BIC:                  -107.3
Df Model:             1
Covariance Type:     nonrobust

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.5035	0.016	32.091	0.000	0.472	0.535
x1	0.0105	0.058	0.182	0.856	-0.105	0.126
Omnibus:	4.976		Durbin-Watson:	0.668		
Prob(Omnibus):	0.083		Jarque-Bera (JB):	4.923		
Skew:	0.698		Prob(JB):	0.0853		
Kurtosis:	2.703		Cond. No.	4.97		

Notes:

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

OLS Regression Results

```

Dep. Variable:        y    R-squared:                0.082
Model:                OLS    Adj. R-squared:         0.067
Method:               Least Squares    F-statistic:         5.549
Date:                Wed, 29 Jun 2022    Prob (F-statistic):   0.0217
Time:                01:05:24    Log-Likelihood:       74.453
No. Observations:    64          AIC:                  -144.9
Df Residuals:        62          BIC:                  -140.6
Df Model:             1
Covariance Type:     nonrobust

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.5115	0.013	38.064	0.000	0.485	0.538
x1	0.1463	0.062	2.356	0.022	0.022	0.270
Omnibus:	1.407		Durbin-Watson:	0.718		
Prob(Omnibus):	0.495		Jarque-Bera (JB):	1.123		
Skew:	-0.324		Prob(JB):	0.570		
Kurtosis:	2.975		Cond. No.	6.62		

Notes:

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

Observamos los mismos problemas.

De esto, podemos concluir que este modelo no es el más adecuado, por lo que ahora se intentará con el resto de medidas de esfuerzo respectivamente para *hitters* y *pitchers*.

Contrato

Análogamente, veámoslo para el valor del contrato del agente libre:

In [18]:

```
# Modelos
hitter_model_I_b = [None]*period
pitcher_model_I_b = [None]*period
```

In [72]:

```
for i in range(1,period):
    # Hitters:
    y_hitter = df_hitter_copy[i]['Promedio_victorias'].tolist()
    x_hitter = df_hitter_copy[i]['Valor_contrato_estandarizado'].tolist()

    x_hitter = sm.add_constant(x_hitter)

    hitter_model_I_b[i] = sm.OLS(y_hitter, x_hitter).fit()

    # Pitchers:
    y_pitcher = df_pitcher_copy[i]['Promedio_victorias'].tolist()
    x_pitcher = df_pitcher_copy[i]['Valor_contrato_estandarizado'].tolist()

    x_pitcher = sm.add_constant(x_pitcher)

    pitcher_model_I_b[i] = sm.OLS(y_pitcher, x_pitcher).fit()
```

Pitchers

In [73]:

```
for i in range(1,period):
    print(pitcher_model_I_b[i].summary())
```

```
=====
                        OLS Regression Results
=====
Dep. Variable:              y    R-squared:                0.068
```

```

Model:                OLS      Adj. R-squared:      0.013
Method:              Least Squares  F-statistic:      1.235
Date:                Wed, 29 Jun 2022  Prob (F-statistic): 0.282
Time:                11:39:21   Log-Likelihood:   25.786
No. Observations:    19        AIC:                -47.57
Df Residuals:        17        BIC:                -45.68
Df Model:            1
Covariance Type:     nonrobust

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.5390	0.018	29.302	0.000	0.500	0.578
x1	-0.0598	0.054	-1.111	0.282	-0.173	0.054
Omnibus:	2.153		Durbin-Watson:	1.363		
Prob(Omnibus):	0.341		Jarque-Bera (JB):	1.583		
Skew:	-0.528		Prob(JB):	0.453		
Kurtosis:	2.059		Cond. No.	3.71		

Notes:

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

OLS Regression Results

```

Dep. Variable:        y      R-squared:      0.014
Model:                OLS      Adj. R-squared:  -0.006
Method:              Least Squares  F-statistic: 0.6884
Date:                Wed, 29 Jun 2022  Prob (F-statistic): 0.411
Time:                11:39:21   Log-Likelihood: 63.904
No. Observations:    51        AIC:                -123.8
Df Residuals:        49        BIC:                -119.9
Df Model:            1
Covariance Type:     nonrobust

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.4964	0.011	44.882	0.000	0.474	0.519
x1	0.0508	0.061	0.830	0.411	-0.072	0.174
Omnibus:	3.690		Durbin-Watson:	0.634		
Prob(Omnibus):	0.158		Jarque-Bera (JB):	2.989		
Skew:	-0.473		Prob(JB):	0.224		
Kurtosis:	2.285		Cond. No.	6.25		

Notes:

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

OLS Regression Results

Dep. Variable:	y	R-squared:	0.002
Model:	OLS	Adj. R-squared:	-0.013
Method:	Least Squares	F-statistic:	0.1443
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.705
Time:	11:39:21	Log-Likelihood:	95.352
No. Observations:	68	AIC:	-186.7
Df Residuals:	66	BIC:	-182.3
Df Model:	1		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
const	0.4886	0.009	51.639	0.000	0.470	0.508
x1	0.0129	0.034	0.380	0.705	-0.055	0.080

Omnibus:	14.508	Durbin-Watson:	1.114
Prob(Omnibus):	0.001	Jarque-Bera (JB):	3.811
Skew:	0.120	Prob(JB):	0.149
Kurtosis:	1.865	Cond. No.	4.77

Notes:

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

OLS Regression Results

Dep. Variable:	y	R-squared:	0.014
Model:	OLS	Adj. R-squared:	-0.001
Method:	Least Squares	F-statistic:	0.9609
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.330
Time:	11:39:21	Log-Likelihood:	92.033
No. Observations:	72	AIC:	-180.1
Df Residuals:	70	BIC:	-175.5
Df Model:	1		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
const	0.5091	0.009	58.468	0.000	0.492	0.526
x1	0.0513	0.052	0.980	0.330	-0.053	0.156

Omnibus:	11.225	Durbin-Watson:	0.718
----------	--------	----------------	-------

Prob(Omnibus):	0.004	Jarque-Bera (JB):	3.997
Skew:	-0.255	Prob(JB):	0.136
Kurtosis:	1.964	Cond. No.	6.53

Notes:

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

OLS Regression Results

Dep. Variable:	y	R-squared:	0.009
Model:	OLS	Adj. R-squared:	-0.004
Method:	Least Squares	F-statistic:	0.6815
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.412
Time:	11:39:21	Log-Likelihood:	104.42
No. Observations:	75	AIC:	-204.8
Df Residuals:	73	BIC:	-200.2
Df Model:	1		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
const	0.5069	0.008	64.677	0.000	0.491	0.523
x1	0.0309	0.037	0.826	0.412	-0.044	0.105

Omnibus:	7.956	Durbin-Watson:	0.587
Prob(Omnibus):	0.019	Jarque-Bera (JB):	2.955
Skew:	-0.089	Prob(JB):	0.228
Kurtosis:	2.044	Cond. No.	5.36

Notes:

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

OLS Regression Results

Dep. Variable:	y	R-squared:	0.063
Model:	OLS	Adj. R-squared:	0.039
Method:	Least Squares	F-statistic:	2.626
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.113
Time:	11:39:21	Log-Likelihood:	55.351
No. Observations:	41	AIC:	-106.7
Df Residuals:	39	BIC:	-103.3
Df Model:	1		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
--	------	---------	---	------	--------	--------

const	0.4959	0.012	41.219	0.000	0.472	0.520
x1	0.0696	0.043	1.621	0.113	-0.017	0.156
=====						
Omnibus:		0.989	Durbin-Watson:			1.632
Prob(Omnibus):		0.610	Jarque-Bera (JB):			0.873
Skew:		0.090	Prob(JB):			0.646
Kurtosis:		2.308	Cond. No.			4.38
=====						

Notes:

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

OLS Regression Results

Dep. Variable:	y	R-squared:	0.001			
Model:	OLS	Adj. R-squared:	-0.018			
Method:	Least Squares	F-statistic:	0.02789			
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.868			
Time:	11:39:21	Log-Likelihood:	59.428			
No. Observations:	55	AIC:	-114.9			
Df Residuals:	53	BIC:	-110.8			
Df Model:	1					
Covariance Type:	nonrobust					
=====						
	coef	std err	t	P> t	[0.025	0.975]

const	0.5223	0.013	38.708	0.000	0.495	0.549
x1	-0.0113	0.067	-0.167	0.868	-0.147	0.124
=====						
Omnibus:	8.274	Durbin-Watson:	0.760			
Prob(Omnibus):	0.016	Jarque-Bera (JB):	8.061			
Skew:	-0.932	Prob(JB):	0.0178			
Kurtosis:	3.216	Cond. No.	6.06			
=====						

Notes:

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

OLS Regression Results

Dep. Variable:	y	R-squared:	0.026
Model:	OLS	Adj. R-squared:	0.005
Method:	Least Squares	F-statistic:	1.220
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.275
Time:	11:39:21	Log-Likelihood:	52.013
No. Observations:	48	AIC:	-100.0

Df Residuals: 46 BIC: -96.28
 Df Model: 1
 Covariance Type: nonrobust

	coef	std err	t	P> t	[0.025	0.975]
const	0.5248	0.014	36.331	0.000	0.496	0.554
x1	0.0812	0.073	1.105	0.275	-0.067	0.229
Omnibus:	7.907		Durbin-Watson:	1.248		
Prob(Omnibus):	0.019		Jarque-Bera (JB):	6.908		
Skew:	-0.854		Prob(JB):	0.0316		
Kurtosis:	3.733		Cond. No.	6.16		

Notes:

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

OLS Regression Results

Dep. Variable: y R-squared: 0.000
 Model: OLS Adj. R-squared: -0.018
 Method: Least Squares F-statistic: 0.0005237
 Date: Wed, 29 Jun 2022 Prob (F-statistic): 0.982
 Time: 11:39:21 Log-Likelihood: 57.704
 No. Observations: 58 AIC: -111.4
 Df Residuals: 56 BIC: -107.3
 Df Model: 1
 Covariance Type: nonrobust

	coef	std err	t	P> t	[0.025	0.975]
const	0.5054	0.013	38.995	0.000	0.479	0.531
x1	-0.0016	0.071	-0.023	0.982	-0.143	0.140
Omnibus:	4.905		Durbin-Watson:	0.668		
Prob(Omnibus):	0.086		Jarque-Bera (JB):	4.859		
Skew:	0.689		Prob(JB):	0.0881		
Kurtosis:	2.668		Cond. No.	5.95		

Notes:

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

OLS Regression Results

Dep. Variable: y R-squared: 0.071

```

Model:                                OLS      Adj. R-squared:      0.056
Method:                               Least Squares      F-statistic:      4.768
Date:                                Wed, 29 Jun 2022      Prob (F-statistic): 0.0328
Time:                                11:39:21      Log-Likelihood:    74.081
No. Observations:                     64      AIC:              -144.2
Df Residuals:                         62      BIC:              -139.8
Df Model:                             1
Covariance Type:                      nonrobust

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.5221	0.011	47.451	0.000	0.500	0.544
x1	0.1461	0.067	2.183	0.033	0.012	0.280

Omnibus:	2.590	Durbin-Watson:	0.730
Prob(Omnibus):	0.274	Jarque-Bera (JB):	2.023
Skew:	-0.431	Prob(JB):	0.364
Kurtosis:	3.126	Cond. No.	6.97

Notes:

Hitters

```
In [74]: for i in range(1,period):
          print(hitter_model_I_b[i].summary())
```

OLS Regression Results

Dep. Variable:	y	R-squared:	0.001
Model:	OLS	Adj. R-squared:	-0.020
Method:	Least Squares	F-statistic:	0.05244
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.820
Time:	11:39:40	Log-Likelihood:	65.642
No. Observations:	49	AIC:	-127.3
Df Residuals:	47	BIC:	-123.5
Df Model:	1		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
const	0.5197	0.010	51.845	0.000	0.499	0.540
x1	-0.0107	0.047	-0.229	0.820	-0.105	0.084

Omnibus:	4.290	Durbin-Watson:	1.360
Prob(Omnibus):	0.117	Jarque-Bera (JB):	4.051
Skew:	-0.649	Prob(JB):	0.132
Kurtosis:	2.452	Cond. No.	5.11

Notes:

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

OLS Regression Results

Dep. Variable:	y	R-squared:	0.011
Model:	OLS	Adj. R-squared:	-0.002
Method:	Least Squares	F-statistic:	0.8739
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.353
Time:	11:39:40	Log-Likelihood:	96.885
No. Observations:	78	AIC:	-189.8
Df Residuals:	76	BIC:	-185.1
Df Model:	1		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
const	0.5063	0.009	55.807	0.000	0.488	0.524
x1	0.0440	0.047	0.935	0.353	-0.050	0.138

Omnibus:	7.944	Durbin-Watson:	0.563
Prob(Omnibus):	0.019	Jarque-Bera (JB):	5.427
Skew:	-0.502	Prob(JB):	0.0663
Kurtosis:	2.187	Cond. No.	5.93

Notes:

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

OLS Regression Results

Dep. Variable:	y	R-squared:	0.002
Model:	OLS	Adj. R-squared:	-0.009
Method:	Least Squares	F-statistic:	0.1657
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.685
Time:	11:39:40	Log-Likelihood:	134.43
No. Observations:	92	AIC:	-264.9
Df Residuals:	90	BIC:	-259.8
Df Model:	1		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
-----	-----	-----	-----	-----	-----	-----
const	0.5042	0.007	77.206	0.000	0.491	0.517
x1	-0.0171	0.042	-0.407	0.685	-0.100	0.066
=====	=====	=====	=====	=====	=====	=====
Omnibus:		12.546	Durbin-Watson:			0.727
Prob(Omnibus):		0.002	Jarque-Bera (JB):			3.855
Skew:		-0.029	Prob(JB):			0.145
Kurtosis:		1.999	Cond. No.			7.12
=====	=====	=====	=====	=====	=====	=====

Notes:

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

OLS Regression Results

Dep. Variable:	y	R-squared:	0.001
Model:	OLS	Adj. R-squared:	-0.012
Method:	Least Squares	F-statistic:	0.06194
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.804
Time:	11:39:41	Log-Likelihood:	101.07
No. Observations:	79	AIC:	-198.1
Df Residuals:	77	BIC:	-193.4
Df Model:	1		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
-----	-----	-----	-----	-----	-----	-----
const	0.5170	0.009	60.637	0.000	0.500	0.534
x1	0.0116	0.047	0.249	0.804	-0.082	0.105
=====	=====	=====	=====	=====	=====	=====
Omnibus:		13.716	Durbin-Watson:			0.685
Prob(Omnibus):		0.001	Jarque-Bera (JB):			5.031
Skew:		-0.332	Prob(JB):			0.0808
Kurtosis:		1.958	Cond. No.			6.14
=====	=====	=====	=====	=====	=====	=====

Notes:

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

OLS Regression Results

Dep. Variable:	y	R-squared:	0.025
Model:	OLS	Adj. R-squared:	0.016
Method:	Least Squares	F-statistic:	2.766
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.0992
Time:	11:39:41	Log-Likelihood:	155.09

```

No. Observations:      108   AIC:                -306.2
Df Residuals:          106   BIC:                -300.8
Df Model:               1
Covariance Type:       nonrobust

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.5149	0.006	83.023	0.000	0.503	0.527
x1	0.0472	0.028	1.663	0.099	-0.009	0.103
Omnibus:	1.555		Durbin-Watson:	0.538		
Prob(Omnibus):	0.460		Jarque-Bera (JB):	1.395		
Skew:	-0.140		Prob(JB):	0.498		
Kurtosis:	2.518		Cond. No.	5.12		

Notes:

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

OLS Regression Results

```

Dep. Variable:          y   R-squared:                0.025
Model:                  OLS   Adj. R-squared:            0.010
Method:                 Least Squares   F-statistic:              1.604
Date:                   Wed, 29 Jun 2022   Prob (F-statistic):       0.210
Time:                   11:39:41   Log-Likelihood:          83.121
No. Observations:       64   AIC:                    -162.2
Df Residuals:           62   BIC:                    -157.9
Df Model:                1
Covariance Type:       nonrobust

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.5072	0.010	49.615	0.000	0.487	0.528
x1	0.0521	0.041	1.267	0.210	-0.030	0.134
Omnibus:	7.056		Durbin-Watson:	0.587		
Prob(Omnibus):	0.029		Jarque-Bera (JB):	3.691		
Skew:	0.366		Prob(JB):	0.158		
Kurtosis:	2.079		Cond. No.	5.01		

Notes:

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

OLS Regression Results

```

Dep. Variable:          y      R-squared:          0.004
Model:                OLS      Adj. R-squared:       -0.014
Method:              Least Squares  F-statistic:        0.2323
Date:                Wed, 29 Jun 2022  Prob (F-statistic):    0.632
Time:                11:39:41    Log-Likelihood:      64.318
No. Observations:      58      AIC:                -124.6
Df Residuals:          56      BIC:                -120.5
Df Model:              1
Covariance Type:      nonrobust

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.5279	0.012	42.438	0.000	0.503	0.553
x1	-0.0239	0.050	-0.482	0.632	-0.123	0.075
Omnibus:	7.645		Durbin-Watson:	0.875		
Prob(Omnibus):	0.022		Jarque-Bera (JB):	6.799		
Skew:	-0.741		Prob(JB):	0.0334		
Kurtosis:	3.785		Cond. No.	4.73		

Notes:

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

OLS Regression Results

Dep. Variable:	y	R-squared:	0.014			
Model:	OLS	Adj. R-squared:	-0.002			
Method:	Least Squares	F-statistic:	0.8675			
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.355			
Time:	11:39:41	Log-Likelihood:	74.376			
No. Observations:	65	AIC:	-144.8			
Df Residuals:	63	BIC:	-140.4			
Df Model:	1					
Covariance Type:	nonrobust					
=====						
	coef	std err	t	P> t	[0.025	0.975]

const	0.5494	0.010	52.464	0.000	0.528	0.570
x1	-0.0530	0.057	-0.931	0.355	-0.167	0.061
=====						
Omnibus:	8.837	Durbin-Watson:	0.793			
Prob(Omnibus):	0.012	Jarque-Bera (JB):	8.289			
Skew:	-0.805	Prob(JB):	0.0159			
Kurtosis:	3.685	Cond. No.	5.90			

Notes:

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

OLS Regression Results

Dep. Variable:	y	R-squared:	0.142
Model:	OLS	Adj. R-squared:	0.124
Method:	Least Squares	F-statistic:	7.623
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.00825
Time:	11:39:41	Log-Likelihood:	58.610
No. Observations:	48	AIC:	-113.2
Df Residuals:	46	BIC:	-109.5
Df Model:	1		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
const	0.4816	0.012	38.534	0.000	0.456	0.507
x1	0.1316	0.048	2.761	0.008	0.036	0.227

Omnibus:	4.874	Durbin-Watson:	1.302
Prob(Omnibus):	0.087	Jarque-Bera (JB):	3.954
Skew:	0.591	Prob(JB):	0.138
Kurtosis:	2.239	Cond. No.	4.62

Notes:

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

OLS Regression Results

Dep. Variable:	y	R-squared:	0.024
Model:	OLS	Adj. R-squared:	0.011
Method:	Least Squares	F-statistic:	1.823
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.181
Time:	11:39:41	Log-Likelihood:	90.953
No. Observations:	75	AIC:	-177.9
Df Residuals:	73	BIC:	-173.3
Df Model:	1		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
const	0.5291	0.010	55.108	0.000	0.510	0.548
x1	0.0642	0.048	1.350	0.181	-0.031	0.159

Omnibus:	0.855	Durbin-Watson:	0.700
Prob(Omnibus):	0.652	Jarque-Bera (JB):	0.901
Skew:	-0.121	Prob(JB):	0.637
Kurtosis:	2.521	Cond. No.	5.70

=====

Notes:

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

A diferencia del salario, el valor del contrato tiende a tener un mayor grado de significancia estadística que en el caso del salario, aunque no el suficiente a lo largo de todos los años como para que se considere un modelo adecuado.

II. Salario en función de las medidas de esfuerzo

$$\omega_t = a + b\delta_t + e_t$$

donde δ_t es un vector que contiene todas las estadísticas deportivas. Se aplicará el modelo tanto para agentes libres que sean *hitters* o *pitchers*. Lo que se hará es regresar sobre todas las medidas del esfuerzo con el objetivo de determinar cuáles variables son significativas para el modelo y cuáles no

```
In [51]: df_hitter_copy[1].columns
```

```
Out[51]: Index(['Jugador', 'Valor_contrato', 'Valor_promedio_contrato', 'Posicion',
               'Juegos', 'Porcentaje_juegos', 'At-bats', 'Bateos', 'Home-runs', 'RBI',
               'Porcentaje_bateo', 'OPS', 'Equipo', 'Sueldo', 'Victorias',
               'Juegos_totales', 'Promedio_victorias', 'Sueldo_estandarizado',
               'Valor_contrato_estandarizado'],
              dtype='object')
```

```
In [50]: df_pitcher_copy[1].columns
```

```
Out[50]: Index(['Jugador', 'Valor_contrato', 'Valor_promedio_contrato', 'Posicion',
               'Juegos', 'Juegos_iniciados', 'Inning_pitched', 'Bateos', 'Carreras',
               'Carreras_ganadas', 'Walks', 'Strike-outs', 'Wins', 'Losses', 'Saves',
               'WHIP', 'ERA', 'Equipo', 'Sueldo', 'Victorias', 'Juegos_totales',
               'Promedio_victorias', 'Sueldo_estandarizado',
               'Valor_contrato_estandarizado'],
              dtype='object')
```

```
In [69]: for i in range(1,period):
# Hitters:
scaler_h = StandardScaler()
x_hitter = df_hitter_copy[i].iloc[:, [4,5,6,7,8,9,10,11,14,15,16]]
scaler_h.fit(x_hitter)
scaler_h.transform(x_hitter)
y_hitter = df_hitter_copy[i]['Sueldo_estandarizado']

x_hitter = sm.add_constant(x_hitter)

hitter_model_II_a[i] = sm.OLS(y_hitter, x_hitter).fit()

# Pitchers:
scaler_p = StandardScaler()
x_pitcher = df_pitcher_copy[i].iloc[:, [4,5,6,7,8,9,10,11,12,13,14,15,16,19,20,21]]
scaler_p.fit(x_pitcher)
scaler_p.transform(x_pitcher)
y_pitcher = df_pitcher_copy[i]['Sueldo_estandarizado']

x_pitcher = sm.add_constant(x_pitcher)

pitcher_model_II_a[i] = sm.OLS(y_pitcher, x_pitcher).fit()
```

```
In [68]: # Modelos
hitter_model_II_a = [None]*period
pitcher_model_II_a = [None]*period
```

Lo siguiente es un código sin normalizar:

```
for i in range(1,period):
```

```
# Hitters:
x_hitter = df_hitter_copy[i][['Juegos', 'Porcentaje_juegos', 'At-bats', 'Bateos', 'Home-
runs', 'RBI',
                                'Porcentaje_bateo', 'OPS', 'Victorias', 'Juegos totales',
                                'Promedio_victorias']]
#x_hitter = normalizar(x_hitter)
y_hitter = df_hitter_copy[i]['Sueldo_estandarizado']
```

```

x_hitter = sm.add_constant(x_hitter)

hitter_model_II_a[i] = sm.OLS(y_hitter, x_hitter).fit()

# Pitchers:
x_pitcher = df_pitcher_copy[i][['Juegos', 'Juegos_iniciados', 'Inning_pitched', 'Bateos',
'Carreras',
                                'Carreras_ganadas', 'Walks', 'Strike-outs', 'Wins',
'Losses', 'Saves',
                                'WHIP', 'ERA', 'Victorias', 'Juegos_totales',
'Promedio_victorias']]
#x_pitcher = normalizar(x_pitcher)
y_pitcher = df_pitcher_copy[i]['Sueldo_estandarizado']

x_pitcher = sm.add_constant(x_pitcher)

pitcher_model_II_a[i] = sm.OLS(y_pitcher, x_pitcher).fit()

```

Hitters

```

In [70]: for i in range(1,period):
          print(hitter_model_II_a[i].summary())

```

```

=====
                        OLS Regression Results
=====
Dep. Variable:          Sueldo_estandarizado    R-squared:                0.322
Model:                  OLS                    Adj. R-squared:           0.166
Method:                 Least Squares          F-statistic:             2.060
Date:                  Wed, 29 Jun 2022        Prob (F-statistic):      0.0580
Time:                  11:03:20                Log-Likelihood:          19.582
No. Observations:      49                     AIC:                    -19.16
Df Residuals:          39                     BIC:                    -0.2454
Df Model:              9
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
Juegos	-0.0009	0.002	-0.551	0.585	-0.004	0.002
Porcetrnaje_juegos	-0.1882	0.369	-0.510	0.613	-0.935	0.559
At-bats	-0.0004	0.001	-0.372	0.712	-0.002	0.002
Bateos	0.0033	0.003	1.051	0.300	-0.003	0.010
Home-runs	-0.0059	0.014	-0.436	0.665	-0.033	0.022

RBI	0.0053	0.006	0.945	0.351	-0.006	0.017
Porcentaje_bateo	-1.0350	1.215	-0.852	0.399	-3.492	1.422
OPS	0.0799	0.399	0.200	0.842	-0.728	0.887
Victorias	0.0009	0.003	0.328	0.744	-0.005	0.007
Juegos totales	0.0012	0.002	0.712	0.480	-0.002	0.005
Promedio_victorias	5.63e-06	1.71e-05	0.328	0.744	-2.9e-05	4.03e-05

```
=====
Omnibus:                12.135    Durbin-Watson:                1.497
Prob(Omnibus):           0.002    Jarque-Bera (JB):           12.489
Skew:                    1.035    Prob(JB):                   0.00194
Kurtosis:                4.354    Cond. No.                   2.73e+18
=====
```

Notes:

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

[2] The smallest eigenvalue is 1.12e-30. This might indicate that there are strong multicollinearity problems or that the design matrix is singular.

OLS Regression Results

```
=====
Dep. Variable:    Sueldo_estandarizado    R-squared:                0.219
Model:            OLS                    Adj. R-squared:           0.116
Method:           Least Squares          F-statistic:             2.121
Date:             Wed, 29 Jun 2022        Prob (F-statistic):      0.0393
Time:             11:03:20                Log-Likelihood:          8.6717
No. Observations: 78                     AIC:                    2.657
Df Residuals:     68                     BIC:                    26.22
Df Model:         9
Covariance Type:  nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
Juegos	-0.1980	0.152	-1.299	0.198	-0.502	0.106
Porcetrnaje_juegos	31.4561	24.717	1.273	0.207	-17.867	80.779
At-bats	0.0012	0.001	1.035	0.305	-0.001	0.004
Bateos	-0.0013	0.004	-0.357	0.722	-0.009	0.006
Home-runs	-0.0041	0.012	-0.332	0.741	-0.029	0.021
RBI	0.0055	0.005	1.128	0.263	-0.004	0.015
Porcentaje_bateo	0.7859	0.819	0.960	0.340	-0.848	2.419
OPS	-0.5325	0.371	-1.437	0.155	-1.272	0.207
Victorias	0.0047	0.003	1.795	0.077	-0.001	0.010
Juegos totales	-0.0002	0.001	-0.171	0.865	-0.003	0.002
Promedio_victorias	2.922e-05	1.63e-05	1.795	0.077	-3.27e-06	6.17e-05

```
=====
Omnibus:                18.966    Durbin-Watson:                1.598
Prob(Omnibus):           0.000    Jarque-Bera (JB):           26.022
=====
```

Skew:	1.047	Prob(JB):	2.24e-06
Kurtosis:	4.904	Cond. No.	4.47e+18

Notes:

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

[2] The smallest eigenvalue is 4.47e-31. This might indicate that there are strong multicollinearity problems or that the design matrix is singular.

OLS Regression Results

Dep. Variable:	Sueldo_estandarizado	R-squared:	0.317
Model:	OLS	Adj. R-squared:	0.223
Method:	Least Squares	F-statistic:	3.375
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.000725
Time:	11:03:20	Log-Likelihood:	24.428
No. Observations:	92	AIC:	-24.86
Df Residuals:	80	BIC:	5.406
Df Model:	11		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
const	3.7646	71.345	0.053	0.958	-138.217	145.746
Juegos	-0.1141	0.464	-0.246	0.806	-1.037	0.809
Porcetrnaje_juegos	17.9890	75.130	0.239	0.811	-131.524	167.502
At-bats	0.0019	0.001	2.404	0.019	0.000	0.003
Bateos	-0.0012	0.003	-0.373	0.710	-0.007	0.005
Home-runs	0.0152	0.008	1.840	0.069	-0.001	0.032
RBI	-0.0030	0.005	-0.612	0.542	-0.013	0.007
Porcentaje_bateo	1.2974	1.230	1.055	0.295	-1.150	3.744
OPS	-0.9638	0.508	-1.898	0.061	-1.974	0.047
Victorias	-0.0664	0.852	-0.078	0.938	-1.762	1.629
Juegos totales	-0.0211	0.441	-0.048	0.962	-0.899	0.857
Promedio_victorias	10.6975	137.658	0.078	0.938	-263.250	284.645

Omnibus:	11.545	Durbin-Watson:	1.677
Prob(Omnibus):	0.003	Jarque-Bera (JB):	11.874
Skew:	0.822	Prob(JB):	0.00264
Kurtosis:	3.626	Cond. No.	2.55e+06

Notes:

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

[2] The condition number is large, 2.55e+06. This might indicate that there are strong multicollinearity or other numerical problems.

OLS Regression Results

Dep. Variable:	Sueldo_estandarizado	R-squared:	0.538
Model:	OLS	Adj. R-squared:	0.462
Method:	Least Squares	F-statistic:	7.079
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	8.21e-08
Time:	11:03:20	Log-Likelihood:	23.974
No. Observations:	79	AIC:	-23.95
Df Residuals:	67	BIC:	4.485
Df Model:	11		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
const	10.8603	77.810	0.140	0.889	-144.449	166.169
Juegos	-0.5847	0.512	-1.143	0.257	-1.606	0.436
Porcetnaje_juegos	93.6330	82.858	1.130	0.262	-71.751	259.017
At-bats	0.0041	0.001	4.194	0.000	0.002	0.006
Bateos	-0.0042	0.003	-1.201	0.234	-0.011	0.003
Home-runs	0.0177	0.008	2.309	0.024	0.002	0.033
RBI	-0.0073	0.004	-1.714	0.091	-0.016	0.001
Porcentaje_bateo	1.6021	0.795	2.014	0.048	0.014	3.190
OPS	-1.0557	0.334	-3.160	0.002	-1.722	-0.389
Victorias	0.4126	0.945	0.436	0.664	-1.474	2.299
Juegos totales	-0.0660	0.484	-0.136	0.892	-1.032	0.900
Promedio_victorias	-66.0053	152.040	-0.434	0.666	-369.479	237.469

Omnibus:	10.366	Durbin-Watson:	1.755
Prob(Omnibus):	0.006	Jarque-Bera (JB):	10.715
Skew:	0.730	Prob(JB):	0.00471
Kurtosis:	4.059	Cond. No.	2.53e+06

Notes:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
 [2] The condition number is large, 2.53e+06. This might indicate that there are strong multicollinearity or other numerical problems.

OLS Regression Results

Dep. Variable:	Sueldo_estandarizado	R-squared:	0.201
Model:	OLS	Adj. R-squared:	0.109
Method:	Least Squares	F-statistic:	2.190
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.0210
Time:	11:03:20	Log-Likelihood:	41.155
No. Observations:	108	AIC:	-58.31

Df Residuals:	96	BIC:	-26.12			
Df Model:	11					
Covariance Type:	nonrobust					
=====						
	coef	std err	t	P> t	[0.025	0.975]

const	-50.2689	53.647	-0.937	0.351	-156.758	56.220
Juegos	-0.0181	0.104	-0.174	0.862	-0.225	0.189
Porcetnaje_juegos	2.7132	16.885	0.161	0.873	-30.803	36.229
At-bats	0.0011	0.001	1.643	0.104	-0.000	0.002
Bateos	-0.0028	0.003	-1.041	0.301	-0.008	0.003
Home-runs	0.0109	0.007	1.557	0.123	-0.003	0.025
RBI	-0.0017	0.004	-0.427	0.670	-0.010	0.006
Porcentaje_bateo	1.3441	0.593	2.267	0.026	0.167	2.521
OPS	-0.5356	0.211	-2.543	0.013	-0.954	-0.117
Victorias	-0.6318	0.631	-1.002	0.319	-1.884	0.620
Juegos totales	0.3115	0.334	0.933	0.353	-0.351	0.974
Promedio_victorias	102.2839	101.311	1.010	0.315	-98.817	303.385
=====						
Omnibus:	74.925	Durbin-Watson:	1.622			
Prob(Omnibus):	0.000	Jarque-Bera (JB):	454.553			
Skew:	2.299	Prob(JB):	1.97e-99			
Kurtosis:	11.937	Cond. No.	2.04e+06			
=====						

Notes:

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

[2] The condition number is large, 2.04e+06. This might indicate that there are strong multicollinearity or other numerical problems.

OLS Regression Results

Dep. Variable:	Sueldo_estandarizado	R-squared:	0.363
Model:	OLS	Adj. R-squared:	0.243
Method:	Least Squares	F-statistic:	3.017
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.00434
Time:	11:03:20	Log-Likelihood:	21.082
No. Observations:	64	AIC:	-20.16
Df Residuals:	53	BIC:	3.583
Df Model:	10		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
const	22.5230	14.809	1.521	0.134	-7.180	52.226
Juegos	0.8581	0.585	1.468	0.148	-0.315	2.031

Porcetnaje_juegos	-139.3765	94.662	-1.472	0.147	-329.244	50.491
At-bats	-0.0002	0.001	-0.194	0.847	-0.002	0.002
Bateos	0.0047	0.003	1.407	0.165	-0.002	0.011
Home-runs	0.0189	0.011	1.732	0.089	-0.003	0.041
RBI	-0.0034	0.005	-0.697	0.489	-0.013	0.006
Porcentaje_bateo	2.6994	1.345	2.007	0.050	0.002	5.397
OPS	-1.1852	0.510	-2.326	0.024	-2.207	-0.163
Victorias	-0.0794	0.054	-1.476	0.146	-0.187	0.029
Juegos totales	-0.1389	0.092	-1.515	0.136	-0.323	0.045
Promedio_victorias	13.4033	8.813	1.521	0.134	-4.273	31.079

Omnibus:	19.694	Durbin-Watson:	1.474
Prob(Omnibus):	0.000	Jarque-Bera (JB):	27.293
Skew:	1.202	Prob(JB):	1.18e-06
Kurtosis:	5.110	Cond. No.	2.66e+18

Notes:

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

[2] The smallest eigenvalue is 1.03e-30. This might indicate that there are strong multicollinearity problems or that the design matrix is singular.

OLS Regression Results

Dep. Variable:	Sueldo_estandarizado	R-squared:	0.299
Model:	OLS	Adj. R-squared:	0.132
Method:	Least Squares	F-statistic:	1.786
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.0844
Time:	11:03:20	Log-Likelihood:	16.357
No. Observations:	58	AIC:	-8.713
Df Residuals:	46	BIC:	16.01
Df Model:	11		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
-----	-----	-----	-----	-----	-----	-----
const	205.6594	504.594	0.408	0.685	-810.036	1221.355
Juegos	-0.1977	0.239	-0.826	0.413	-0.679	0.284
Porcetnaje_juegos	31.5120	38.806	0.812	0.421	-46.600	109.624
At-bats	-0.0005	0.001	-0.577	0.567	-0.002	0.001
Bateos	0.0048	0.003	1.462	0.151	-0.002	0.012
Home-runs	0.0079	0.015	0.533	0.596	-0.022	0.038
RBI	0.0012	0.006	0.201	0.842	-0.011	0.014
Porcentaje_bateo	0.5193	0.907	0.572	0.570	-1.307	2.345
OPS	-0.2708	0.350	-0.774	0.443	-0.975	0.433
Victorias	2.4694	5.321	0.464	0.645	-8.241	13.180

Juegos totales	-1.2659	3.115	-0.406	0.686	-7.535	5.003
Promedio_victorias	-400.6241	862.021	-0.465	0.644	-2135.782	1334.533
=====						
Omnibus:	38.934	Durbin-Watson:	1.516			
Prob(Omnibus):	0.000	Jarque-Bera (JB):	124.196			
Skew:	1.906	Prob(JB):	1.07e-27			
Kurtosis:	9.071	Cond. No.	1.23e+07			
=====						

Notes:

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

[2] The condition number is large, 1.23e+07. This might indicate that there are strong multicollinearity or other numerical problems.

OLS Regression Results

Dep. Variable:	Sueldo_estandarizado	R-squared:	0.405			
Model:	OLS	Adj. R-squared:	0.281			
Method:	Least Squares	F-statistic:	3.273			
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.00179			
Time:	11:03:20	Log-Likelihood:	16.356			
No. Observations:	65	AIC:	-8.713			
Df Residuals:	53	BIC:	17.38			
Df Model:	11					
Covariance Type:	nonrobust					
=====						
	coef	std err	t	P> t	[0.025	0.975]

const	102.9063	69.455	1.482	0.144	-36.402	242.215
Juegos	0.5705	0.280	2.041	0.046	0.010	1.131
Porcetrnaje_juegos	-92.9827	45.339	-2.051	0.045	-183.921	-2.045
At-bats	-0.0005	0.001	-0.431	0.669	-0.003	0.002
Bateos	0.0066	0.004	1.690	0.097	-0.001	0.014
Home-runs	0.0266	0.007	3.560	0.001	0.012	0.042
RBI	-0.0059	0.004	-1.559	0.125	-0.013	0.002
Porcentaje_bateo	1.3297	1.081	1.230	0.224	-0.839	3.499
OPS	-0.9315	0.492	-1.894	0.064	-1.918	0.055
Victorias	1.0383	0.772	1.345	0.185	-0.511	2.587
Juegos totales	-0.6342	0.430	-1.475	0.146	-1.497	0.228
Promedio_victorias	-167.6584	124.735	-1.344	0.185	-417.845	82.528
=====						
Omnibus:	5.786	Durbin-Watson:	1.778			
Prob(Omnibus):	0.055	Jarque-Bera (JB):	5.027			
Skew:	0.659	Prob(JB):	0.0810			
Kurtosis:	3.343	Cond. No.	2.11e+06			
=====						

Notes:

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

[2] The condition number is large, 2.11e+06. This might indicate that there are strong multicollinearity or other numerical problems.

OLS Regression Results

Dep. Variable:	Sueldo_estandarizado	R-squared:	0.405
Model:	OLS	Adj. R-squared:	0.223
Method:	Least Squares	F-statistic:	2.229
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.0347
Time:	11:03:20	Log-Likelihood:	10.667
No. Observations:	48	AIC:	2.665
Df Residuals:	36	BIC:	25.12
Df Model:	11		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
-----	-----	-----	-----	-----	-----	-----
const	2.2861	13.614	0.168	0.868	-25.324	29.896
Juegos	-0.0487	0.170	-0.286	0.776	-0.394	0.296
Porcetnaje_juegos	2.2530	10.265	0.219	0.828	-18.564	23.070
At-bats	0.0052	0.004	1.172	0.249	-0.004	0.014
Bateos	-0.0036	0.014	-0.262	0.795	-0.032	0.024
Home-runs	0.0010	0.025	0.039	0.969	-0.051	0.053
RBI	0.0021	0.010	0.220	0.827	-0.017	0.022
Porcentaje_bateo	-1.6954	2.272	-0.746	0.460	-6.302	2.912
OPS	0.6447	0.652	0.989	0.329	-0.677	1.966
Victorias	0.1033	0.480	0.215	0.831	-0.869	1.076
Juegos totales	-0.0449	0.227	-0.198	0.845	-0.506	0.416
Promedio_victorias	-5.3127	28.757	-0.185	0.854	-63.635	53.010

Omnibus:	6.538	Durbin-Watson:	1.450
Prob(Omnibus):	0.038	Jarque-Bera (JB):	6.260
Skew:	0.533	Prob(JB):	0.0437
Kurtosis:	4.411	Cond. No.	1.56e+05

Notes:

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

[2] The condition number is large, 1.56e+05. This might indicate that there are strong multicollinearity or other numerical problems.

OLS Regression Results

Dep. Variable:	Sueldo_estandarizado	R-squared:	0.313
----------------	----------------------	------------	-------

```

Model:                                OLS      Adj. R-squared:      0.193
Method:                            Least Squares  F-statistic:      2.609
Date:                               Wed, 29 Jun 2022  Prob (F-statistic): 0.00837
Time:                               11:03:20      Log-Likelihood:   31.151
No. Observations:                   75          AIC:              -38.30
Df Residuals:                       63          BIC:              -10.49
Df Model:                           11
Covariance Type:                    nonrobust

```

	coef	std err	t	P> t	[0.025	0.975]
const	-66.2732	40.882	-1.621	0.110	-147.970	15.424
Juegos	0.0330	0.151	0.219	0.828	-0.269	0.335
Porcetnaje_juegos	-6.1553	24.490	-0.251	0.802	-55.095	42.785
At-bats	0.0009	0.001	1.004	0.319	-0.001	0.003
Bateos	0.0020	0.003	0.695	0.490	-0.004	0.008
Home-runs	-5.796e-05	0.007	-0.009	0.993	-0.013	0.013
RBI	7.221e-05	0.004	0.019	0.985	-0.007	0.008
Porcentaje_bateo	-0.2200	0.942	-0.234	0.816	-2.102	1.662
OPS	-0.0740	0.361	-0.205	0.838	-0.795	0.647
Victorias	-0.7591	0.461	-1.648	0.104	-1.680	0.162
Juegos totales	0.4091	0.252	1.622	0.110	-0.095	0.913
Promedio_victorias	123.6057	74.674	1.655	0.103	-25.619	272.830
Omnibus:	27.830		Durbin-Watson:	1.437		
Prob(Omnibus):	0.000		Jarque-Bera (JB):	51.852		
Skew:	1.350		Prob(JB):	5.50e-12		
Kurtosis:	6.051		Cond. No.	1.54e+06		

Notes:

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

[2] The condition number is large, 1.54e+06. This might indicate that there are strong multicollinearity or other numerical problems

Se aprecia que la variable que tiende a ser estadísticamente significativa a lo largo de los años es *at-bats* y en menos ocasiones, secundándole, es *home-runs* y *victorias*.

Pitcher

```
In [71]: for i in range(1,period):
          print(pitcher_model_II_a[i].summary())
```

OLS Regression Results


```

=====
Dep. Variable:    Sueldo_estandarizado    R-squared:                0.930
Model:            OLS                    Adj. R-squared:           0.685
Method:          Least Squares           F-statistic:              3.797
Date:            Wed, 29 Jun 2022         Prob (F-statistic):       0.103
Time:            11:30:45                Log-Likelihood:           18.467
No. Observations: 19                    AIC:                      -6.934
Df Residuals:    4                      BIC:                      7.233
Df Model:        14
Covariance Type: nonrobust
=====

```

```

=====
              coef      std err          t      P>|t|      [0.025      0.975]
-----
Juegos          -0.0430      0.019      -2.290      0.084      -0.095      0.009
Juegos_iniciados -0.2394      0.106      -2.257      0.087      -0.534      0.055
Inning_pitched  -0.0083      0.033      -0.253      0.813      -0.100      0.083
Bateos           0.0040      0.027       0.148      0.890      -0.071      0.079
Carreras        -0.0359      0.038      -0.948      0.397      -0.141      0.069
Carreras_ganadas 0.0342      0.033       1.027      0.363      -0.058      0.127
Walks            0.0569      0.022       2.597      0.060      -0.004      0.118
Strike-outs      0.0022      0.007       0.299      0.780      -0.018      0.023
Wins             0.2009      0.091       2.209      0.092      -0.052      0.454
Losses           0.3156      0.123       2.567      0.062      -0.026      0.657
Saves           -0.0156      0.013      -1.177      0.304      -0.053      0.021
WHIP            -5.9510      2.602      -2.287      0.084     -13.174      1.272
ERA              0.9809      0.389       2.521      0.065      -0.099      2.061
Victorias        0.0236      0.010       2.393      0.075      -0.004      0.051
Juegos_totales   0.0183      0.011       1.625      0.180      -0.013      0.050
Promedio_victorias 0.0001     6.09e-05      2.393      0.075     -2.34e-05      0.000
=====

```

```

=====
Omnibus:            5.571    Durbin-Watson:           1.917
Prob(Omnibus):      0.062    Jarque-Bera (JB):        3.114
Skew:               0.736    Prob(JB):                0.211
Kurtosis:           4.329    Cond. No.                5.88e+18
=====

```

Notes:

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

[2] The smallest eigenvalue is 4.15e-32. This might indicate that there are strong multicollinearity problems or that the design matrix is singular.

OLS Regression Results

```

=====
Dep. Variable:    Sueldo_estandarizado    R-squared:                0.535
Model:            OLS                    Adj. R-squared:           0.355
Method:          Least Squares           F-statistic:              2.964

```

```

Date:                Wed, 29 Jun 2022    Prob (F-statistic):    0.00432
Time:                11:30:45           Log-Likelihood:       25.572
No. Observations:    51                AIC:                 -21.14
Df Residuals:        36                BIC:                 7.834
Df Model:            14
Covariance Type:     nonrobust

```

	coef	std err	t	P> t	[0.025	0.975]

Juegos	-0.0014	0.004	-0.343	0.733	-0.009	0.007
Juegos_iniciados	0.0082	0.022	0.370	0.713	-0.037	0.053
Inning_pitched	-0.0013	0.005	-0.251	0.803	-0.011	0.009
Bateos	-0.0035	0.004	-0.838	0.408	-0.012	0.005
Carreras	-0.0067	0.015	-0.442	0.661	-0.037	0.024
Carreras_ganadas	0.0064	0.018	0.350	0.728	-0.031	0.043
Walks	-0.0035	0.004	-0.888	0.381	-0.012	0.005
Strike-outs	0.0018	0.002	0.999	0.324	-0.002	0.005
Wins	0.0390	0.020	1.989	0.054	-0.001	0.079
Losses	0.0442	0.019	2.282	0.028	0.005	0.084
Saves	0.0082	0.004	2.263	0.030	0.001	0.015
WHIP	0.0785	0.264	0.298	0.768	-0.456	0.613
ERA	-0.0133	0.040	-0.332	0.742	-0.095	0.068
Victorias	-0.0010	0.003	-0.368	0.715	-0.007	0.005
Juegos_totales	0.0009	0.002	0.435	0.667	-0.003	0.005
Promedio_victorias	-6.29e-06	1.71e-05	-0.368	0.715	-4.1e-05	2.84e-05
=====						
Omnibus:	28.084	Durbin-Watson:		1.600		
Prob(Omnibus):	0.000	Jarque-Bera (JB):		64.811		
Skew:	1.545	Prob(JB):		8.44e-15		
Kurtosis:	7.577	Cond. No.		4.61e+18		
=====						

Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
[2] The smallest eigenvalue is 1.63e-31. This might indicate that there are strong multicollinearity problems or that the design matrix is singular.

OLS Regression Results

=====			
Dep. Variable:	Sueldo_estandarizado	R-squared:	0.510
Model:	OLS	Adj. R-squared:	0.357
Method:	Least Squares	F-statistic:	3.322
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.000558
Time:	11:30:45	Log-Likelihood:	19.277
No. Observations:	68	AIC:	-4.554
Df Residuals:	51	BIC:	33.18
=====			

Df Model:	16					
Covariance Type:	nonrobust					
	coef	std err	t	P> t	[0.025	0.975]
const	43.6971	78.345	0.558	0.579	-113.587	200.981
Juegos	-0.0004	0.002	-0.245	0.807	-0.004	0.003
Juegos_iniciados	0.0028	0.002	1.268	0.211	-0.002	0.007
Inning_pitched	0.0033	0.005	0.701	0.487	-0.006	0.013
Bateos	-0.0011	0.005	-0.236	0.814	-0.010	0.008
Carreras	0.0107	0.011	1.010	0.317	-0.011	0.032
Carreras_ganadas	-0.0084	0.012	-0.697	0.489	-0.033	0.016
Walks	-0.0046	0.004	-1.206	0.233	-0.012	0.003
Strike-outs	0.0013	0.002	0.588	0.559	-0.003	0.006
Wins	-0.0162	0.018	-0.888	0.379	-0.053	0.020
Losses	-0.0036	0.022	-0.165	0.869	-0.047	0.040
Saves	0.0048	0.003	1.487	0.143	-0.002	0.011
WHIP	0.1205	0.132	0.913	0.366	-0.145	0.385
ERA	-0.0026	0.020	-0.131	0.897	-0.042	0.037
Victorias	0.4070	0.963	0.423	0.674	-1.526	2.340
Juegos totales	-0.2717	0.485	-0.560	0.578	-1.246	0.702
Promedio_victorias	-65.5662	155.411	-0.422	0.675	-377.567	246.435
Omnibus:	7.190	Durbin-Watson:		2.092		
Prob(Omnibus):	0.027	Jarque-Bera (JB):		6.562		
Skew:	0.734	Prob(JB):		0.0376		
Kurtosis:	3.400	Cond. No.		1.68e+06		

Notes:

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

[2] The condition number is large, 1.68e+06. This might indicate that there are strong multicollinearity or other numerical problems.

OLS Regression Results

Dep. Variable:	Sueldo_estandarizado	R-squared:	0.707
Model:	OLS	Adj. R-squared:	0.621
Method:	Least Squares	F-statistic:	8.276
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	1.34e-09
Time:	11:30:45	Log-Likelihood:	52.821
No. Observations:	72	AIC:	-71.64
Df Residuals:	55	BIC:	-32.94
Df Model:	16		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]

const	-16.2781	49.290	-0.330	0.742	-115.058	82.502
Juegos	-0.0009	0.001	-1.026	0.309	-0.003	0.001
Juegos_iniciados	0.0028	0.001	2.522	0.015	0.001	0.005
Inning_pitched	-0.0055	0.003	-1.639	0.107	-0.012	0.001
Bateos	0.0061	0.003	1.998	0.051	-1.78e-05	0.012
Carreras	-0.0094	0.011	-0.830	0.410	-0.032	0.013
Carreras_ganadas	0.0031	0.011	0.279	0.781	-0.019	0.026
Walks	-0.0016	0.002	-0.673	0.503	-0.006	0.003
Strike-outs	0.0047	0.001	3.828	0.000	0.002	0.007
Wins	0.0076	0.011	0.675	0.503	-0.015	0.030
Losses	0.0112	0.011	1.006	0.319	-0.011	0.033
Saves	0.0035	0.002	1.484	0.144	-0.001	0.008
WHIP	-0.1753	0.088	-1.983	0.052	-0.352	0.002
ERA	0.0235	0.015	1.594	0.117	-0.006	0.053
Victorias	-0.2304	0.624	-0.369	0.713	-1.480	1.020
Juegos_totales	0.1017	0.307	0.332	0.741	-0.513	0.716
Promedio_victorias	37.2402	100.281	0.371	0.712	-163.728	238.208
=====						
Omnibus:	16.849	Durbin-Watson:	1.748			
Prob(Omnibus):	0.000	Jarque-Bera (JB):	25.200			
Skew:	0.908	Prob(JB):	3.37e-06			
Kurtosis:	5.259	Cond. No.	1.64e+06			
=====						

Notes:

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

[2] The condition number is large, 1.64e+06. This might indicate that there are strong multicollinearity or other numerical problems.

OLS Regression Results

Dep. Variable:	Sueldo_estandarizado	R-squared:	0.649			
Model:	OLS	Adj. R-squared:	0.552			
Method:	Least Squares	F-statistic:	6.692			
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	3.09e-08			
Time:	11:30:45	Log-Likelihood:	59.864			
No. Observations:	75	AIC:	-85.73			
Df Residuals:	58	BIC:	-46.33			
Df Model:	16					
Covariance Type:	nonrobust					
=====						
	coef	std err	t	P> t	[0.025	0.975]

const	-64.7772	50.905	-1.273	0.208	-166.675	37.121

Juegos	0.0012	0.002	0.618	0.539	-0.003	0.005
Juegos_iniciados	0.0050	0.009	0.571	0.570	-0.012	0.022
Inning_pitched	-0.0014	0.004	-0.395	0.695	-0.009	0.006
Bateos	-0.0021	0.003	-0.629	0.532	-0.009	0.005
Carreras	-0.0073	0.009	-0.825	0.413	-0.025	0.010
Carreras_ganadas	0.0165	0.010	1.684	0.097	-0.003	0.036
Walks	-0.0088	0.003	-2.914	0.005	-0.015	-0.003
Strike-outs	0.0023	0.001	1.801	0.077	-0.000	0.005
Wins	0.0212	0.008	2.592	0.012	0.005	0.038
Losses	-0.0016	0.009	-0.182	0.856	-0.019	0.016
Saves	0.0031	0.003	1.007	0.318	-0.003	0.009
WHIP	0.1735	0.195	0.892	0.376	-0.216	0.563
ERA	-0.0279	0.026	-1.069	0.290	-0.080	0.024
Victorias	-0.9399	0.622	-1.510	0.136	-2.186	0.306
Juegos_totales	0.4014	0.317	1.266	0.211	-0.233	1.036
Promedio_victorias	151.3231	100.029	1.513	0.136	-48.907	351.553

Omnibus:	46.886	Durbin-Watson:	1.850
Prob(Omnibus):	0.000	Jarque-Bera (JB):	258.611
Skew:	1.703	Prob(JB):	6.97e-57
Kurtosis:	11.435	Cond. No.	2.00e+06

Notes:

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

[2] The condition number is large, 2e+06. This might indicate that there are strong multicollinearity or other numerical problems.

OLS Regression Results

Dep. Variable:	Sueldo_estandarizado	R-squared:	0.561
Model:	OLS	Adj. R-squared:	0.297
Method:	Least Squares	F-statistic:	2.126
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.0462
Time:	11:30:45	Log-Likelihood:	17.973
No. Observations:	41	AIC:	-3.946
Df Residuals:	25	BIC:	23.47
Df Model:	15		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
-----	-----	-----	-----	-----	-----	-----
const	16.9374	22.369	0.757	0.456	-29.131	63.006
Juegos	-0.0042	0.002	-1.745	0.093	-0.009	0.001
Juegos_iniciados	0.0024	0.002	1.239	0.227	-0.002	0.006
Inning_pitched	0.0031	0.007	0.457	0.652	-0.011	0.017

Bateos	0.0006	0.007	0.085	0.933	-0.013	0.014
Carreras	-0.0151	0.022	-0.677	0.505	-0.061	0.031
Carreras_ganadas	-0.0020	0.019	-0.102	0.920	-0.042	0.038
Walks	-0.0051	0.006	-0.891	0.382	-0.017	0.007
Strike-outs	0.0017	0.003	0.539	0.595	-0.005	0.008
Wins	0.0046	0.024	0.190	0.851	-0.045	0.054
Losses	0.0600	0.022	2.735	0.011	0.015	0.105
Saves	0.0056	0.004	1.528	0.139	-0.002	0.013
WHIP	0.1223	0.145	0.846	0.405	-0.175	0.420
ERA	-0.0165	0.014	-1.162	0.256	-0.046	0.013
Victorias	-0.0637	0.080	-0.800	0.431	-0.228	0.100
Juegos totales	-0.1007	0.139	-0.724	0.476	-0.387	0.186
Promedio_victorias	10.0793	13.311	0.757	0.456	-17.336	37.495

Omnibus:	4.743	Durbin-Watson:	1.556
Prob(Omnibus):	0.093	Jarque-Bera (JB):	3.443
Skew:	0.565	Prob(JB):	0.179
Kurtosis:	3.860	Cond. No.	1.75e+18

Notes:

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

[2] The smallest eigenvalue is 8.21e-31. This might indicate that there are strong multicollinearity problems or that the design matrix is singular.

OLS Regression Results

Dep. Variable:	Sueldo_estandarizado	R-squared:	0.300
Model:	OLS	Adj. R-squared:	0.006
Method:	Least Squares	F-statistic:	1.019
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.459
Time:	11:30:45	Log-Likelihood:	23.375
No. Observations:	55	AIC:	-12.75
Df Residuals:	38	BIC:	21.37
Df Model:	16		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
-----	-----	-----	-----	-----	-----	-----
const	-276.1550	438.327	-0.630	0.532	-1163.502	611.192
Juegos	-0.0017	0.002	-1.074	0.289	-0.005	0.001
Juegos_iniciados	0.0024	0.003	0.905	0.371	-0.003	0.008
Inning_pitched	-0.0024	0.005	-0.443	0.660	-0.013	0.008
Bateos	0.0014	0.005	0.266	0.791	-0.009	0.012
Carreras	0.0137	0.011	1.285	0.207	-0.008	0.035
Carreras_ganadas	-0.0147	0.011	-1.323	0.194	-0.037	0.008

Walks	-0.0001	0.004	-0.038	0.970	-0.008	0.008
Strike-outs	0.0008	0.003	0.261	0.796	-0.005	0.007
Wins	0.0069	0.022	0.315	0.755	-0.038	0.052
Losses	0.0148	0.023	0.630	0.533	-0.033	0.062
Saves	0.0043	0.004	1.154	0.256	-0.003	0.012
WHIP	-0.0525	0.166	-0.316	0.754	-0.389	0.284
ERA	0.0067	0.023	0.294	0.770	-0.040	0.053
Victorias	-2.7945	4.683	-0.597	0.554	-12.275	6.686
Juegos totales	1.7052	2.706	0.630	0.532	-3.772	7.182
Promedio_victorias	452.8614	758.754	0.597	0.554	-1083.156	1988.879

Omnibus:	25.666	Durbin-Watson:	1.545
Prob(Omnibus):	0.000	Jarque-Bera (JB):	51.791
Skew:	1.415	Prob(JB):	5.67e-12
Kurtosis:	6.820	Cond. No.	8.09e+06

Notes:

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

[2] The condition number is large, 8.09e+06. This might indicate that there are strong multicollinearity or other numerical problems.

OLS Regression Results

Dep. Variable:	Sueldo_estandarizado	R-squared:	0.286
Model:	OLS	Adj. R-squared:	-0.082
Method:	Least Squares	F-statistic:	0.7768
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.698
Time:	11:30:45	Log-Likelihood:	8.4654
No. Observations:	48	AIC:	17.07
Df Residuals:	31	BIC:	48.88
Df Model:	16		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
-----	-----	-----	-----	-----	-----	-----
const	36.7573	96.420	0.381	0.706	-159.892	233.407
Juegos	-0.0034	0.002	-1.535	0.135	-0.008	0.001
Juegos_iniciados	0.0041	0.003	1.404	0.170	-0.002	0.010
Inning_pitched	-0.0007	0.008	-0.091	0.928	-0.016	0.015
Bateos	-0.0053	0.008	-0.656	0.517	-0.022	0.011
Carreras	-0.0354	0.021	-1.662	0.107	-0.079	0.008
Carreras_ganadas	0.0392	0.022	1.814	0.079	-0.005	0.083
Walks	0.0077	0.008	0.981	0.334	-0.008	0.024
Strike-outs	0.0018	0.003	0.691	0.495	-0.003	0.007
Wins	0.0150	0.025	0.604	0.550	-0.036	0.066

Losses	0.0149	0.024	0.619	0.541	-0.034	0.064
Saves	0.0068	0.012	0.583	0.564	-0.017	0.030
WHIP	0.0199	0.097	0.205	0.839	-0.178	0.218
ERA	-0.0021	0.015	-0.142	0.888	-0.033	0.028
Victorias	0.4332	1.122	0.386	0.702	-1.855	2.722
Juegos totales	-0.2277	0.598	-0.381	0.706	-1.447	0.992
Promedio_victorias	-69.5376	180.959	-0.384	0.703	-438.607	299.532

Omnibus:	7.602	Durbin-Watson:	1.693
Prob(Omnibus):	0.022	Jarque-Bera (JB):	6.710
Skew:	0.733	Prob(JB):	0.0349
Kurtosis:	4.099	Cond. No.	1.37e+06

Notes:

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

[2] The condition number is large, 1.37e+06. This might indicate that there are strong multicollinearity or other numerical problems.

OLS Regression Results

Dep. Variable:	Sueldo_estandarizado	R-squared:	0.534
Model:	OLS	Adj. R-squared:	0.353
Method:	Least Squares	F-statistic:	2.940
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.00273
Time:	11:30:45	Log-Likelihood:	30.124
No. Observations:	58	AIC:	-26.25
Df Residuals:	41	BIC:	8.781
Df Model:	16		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
-----	-----	-----	-----	-----	-----	-----
const	-5.7843	9.224	-0.627	0.534	-24.412	12.844
Juegos	-0.0072	0.004	-1.958	0.057	-0.015	0.000
Juegos_iniciados	0.0066	0.004	1.764	0.085	-0.001	0.014
Inning_pitched	0.0024	0.009	0.270	0.789	-0.015	0.020
Bateos	0.0063	0.008	0.767	0.448	-0.010	0.023
Carreras	0.0104	0.022	0.473	0.638	-0.034	0.055
Carreras_ganadas	-0.0270	0.022	-1.227	0.227	-0.071	0.017
Walks	0.0023	0.008	0.275	0.785	-0.015	0.019
Strike-outs	0.0023	0.003	0.709	0.482	-0.004	0.009
Wins	0.0024	0.029	0.084	0.934	-0.056	0.061
Losses	-0.0197	0.028	-0.696	0.490	-0.077	0.038
Saves	0.0033	0.012	0.272	0.787	-0.021	0.028
WHIP	-0.2415	0.094	-2.580	0.014	-0.431	-0.052

ERA	0.0656	0.022	3.032	0.004	0.022	0.109
Victorias	-0.1139	0.316	-0.361	0.720	-0.751	0.524
Juegos totales	0.0990	0.155	0.640	0.526	-0.213	0.411
Promedio_victorias	6.9038	18.820	0.367	0.716	-31.103	44.911

Omnibus:	14.765	Durbin-Watson:	1.532
Prob(Omnibus):	0.001	Jarque-Bera (JB):	21.786
Skew:	0.877	Prob(JB):	1.86e-05
Kurtosis:	5.437	Cond. No.	8.35e+04

Notes:

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

[2] The condition number is large, 8.35e+04. This might indicate that there are strong multicollinearity or other numerical problems.

OLS Regression Results

Dep. Variable:	Sueldo_estandarizado	R-squared:	0.498
Model:	OLS	Adj. R-squared:	0.327
Method:	Least Squares	F-statistic:	2.911
Date:	Wed, 29 Jun 2022	Prob (F-statistic):	0.00225
Time:	11:30:45	Log-Likelihood:	49.512
No. Observations:	64	AIC:	-65.02
Df Residuals:	47	BIC:	-28.32
Df Model:	16		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
-----	-----	-----	-----	-----	-----	-----
const	-52.2350	29.254	-1.786	0.081	-111.087	6.617
Juegos	-0.0021	0.001	-2.495	0.016	-0.004	-0.000
Juegos_iniciados	0.0025	0.001	2.723	0.009	0.001	0.004
Inning_pitched	-0.0009	0.003	-0.296	0.768	-0.007	0.005
Bateos	-0.0054	0.003	-1.803	0.078	-0.011	0.001
Carreras	0.0061	0.009	0.682	0.499	-0.012	0.024
Carreras_ganadas	-0.0004	0.008	-0.050	0.960	-0.017	0.017
Walks	0.0038	0.003	1.166	0.250	-0.003	0.010
Strike-outs	0.0009	0.001	0.763	0.449	-0.002	0.003
Wins	0.0208	0.009	2.208	0.032	0.002	0.040
Losses	0.0081	0.011	0.753	0.455	-0.013	0.030
Saves	0.0040	0.003	1.556	0.126	-0.001	0.009
WHIP	0.0373	0.067	0.559	0.579	-0.097	0.171
ERA	-0.0092	0.008	-1.092	0.280	-0.026	0.008
Victorias	-0.6734	0.342	-1.970	0.055	-1.361	0.014
Juegos_totales	0.3209	0.180	1.781	0.081	-0.042	0.683

Promedio_victorias	109.7434	55.525	1.976	0.054	-1.959	221.446
=====						
Omnibus:	45.549	Durbin-Watson:	1.436			
Prob(Omnibus):	0.000	Jarque-Bera (JB):	318.257			
Skew:	1.719	Prob(JB):	7.79e-70			
Kurtosis:	13.370	Cond. No.	9.18e+05			
=====						

Notes:

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

[2] The condition number is large, 9.18e+05. This might indicate that there are strong multicollinearity or other numerical problems.

Notamos que no hay alguna variable que persista más de tres años siendo estadísticamente significativa. EL comportamiento tiende a ser aleatorio.