

## modelacion\_mexico

November 18, 2025

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[8]: # Limpieza de datos

import pandas as pd

df = pd.read_csv('../data/raw/datos_mexico.csv')

df=df.drop(['_id', 'REGLON', 'ENTIDAD'],axis=1)

# Combinacion de columnas

df['Poblacion de jovenes (3-14)']=df['POB_3_5']+df['POB_6_011']+df['POB_012_014']
df=df.drop(['POB_3_5', 'POB_6_011', 'POB_012_014'],axis=1)
df['Poblacion de adultos (15-64)']=df['POB_015_29']+df['POB_30_64']
df = df.
    ↳drop(['POB_015_017', 'POB_015_019', 'POB_015_29', 'POB_015_49', 'POB_018_24', 'POB_20_24', 'POB_30_
df = df.rename(columns={'POB_65_MAS':'Poblacion adultos mayores (65+)'})

# Calculo de tasas de cambio por año

df['Cambio jovenes']=df['Poblacion de jovenes (3-14)'].diff().fillna(0)
df['Cambio adultos']=df['Poblacion de adultos (15-64)'].diff().fillna(0)
df['Cambio mayores']=df['Poblacion adultos mayores (65+)'].diff().fillna(0)

print(df)
```

	ANIO	Poblacion adultos mayores (65+)	Poblacion de jovenes (3-14)	\
0	2002	5399307	27773497	
1	2003	5576695	27829732	
2	2004	5761262	27861649	
3	2005	5953231	27858970	
4	2006	6155725	27830965	
5	2007	6368873	27795480	
6	2008	6585485	27758562	
7	2009	6808607	27724088	
8	2010	7046284	27705408	
9	2011	7302545	27706657	
10	2012	7569447	27709468	
11	2013	7840039	27692281	

12	2014	8110040	27630432
13	2015	8375584	27511332
14	2017	8935919	27218511
15	2018	9257526	27083534
16	2019	9602115	26933238
17	2020	9860625	26754801
18	2021	10027670	26563170
19	2022	10325860	26360126

	Poblacion de adultos (15-64)	Cambio jovenes	Cambio adultos \
0	61886631	0.0	0.0
1	63131727	56235.0	1245096.0
2	64366713	31917.0	1234986.0
3	65630336	-2679.0	1263623.0
4	66991182	-28005.0	1360846.0
5	68456495	-35485.0	1465313.0
6	69986944	-36918.0	1530449.0
7	71525650	-34474.0	1538706.0
8	73022807	-18680.0	1497157.0
9	74476574	1249.0	1453767.0
10	75877476	2811.0	1400902.0
11	77236926	-17187.0	1359450.0
12	78555813	-61849.0	1318887.0
13	79803346	-119100.0	1247533.0
14	82127176	-292821.0	2323830.0
15	83247972	-134977.0	1120796.0
16	84329698	-150296.0	1081726.0
17	85272128	-178437.0	942430.0
18	86098570	-191631.0	826442.0
19	87023345	-203044.0	924775.0

	Cambio mayores
0	0.0
1	177388.0
2	184567.0
3	191969.0
4	202494.0
5	213148.0
6	216612.0
7	223122.0
8	237677.0
9	256261.0
10	266902.0
11	270592.0
12	270001.0
13	265544.0
14	560335.0
15	321607.0

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16      344589.0
17      258510.0
18      167045.0
19      298190.0

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[ ]: #df.to_csv('../data/mexico.csv')
```

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[ ]: # Modelacion estadistica
import statsmodels.api as sm

# Asegura float en columnas usadas
cols = [
    "Poblacion de adultos (15-64)",
    "Poblacion de jovenes (3-14)",
    "Poblacion adultos mayores (65+)",
    "Cambio jovenes",
    "Cambio adultos",
    "Cambio mayores",
]
df[cols] = df[cols].apply(pd.to_numeric, errors="coerce")
df = df.dropna(subset=cols)

x_J = df[["Poblacion de adultos (15-64)", "Poblacion de jovenes (3-14)"]]
y_J = df["Cambio jovenes"]
# x_J = sm.add_constant(x_J,prepend=False)
modelo_J = sm.OLS(
    y_J,
    x_J,
).fit()
coef_J = modelo_J.params.to_dict()

x_A = df[["Poblacion de jovenes (3-14)", "Poblacion de adultos (15-64)"]]
y_A = df["Cambio adultos"]
# x_A = sm.add_constant(x_A,prepend=False)
modelo_A = sm.OLS(
    y_A,
    x_A,
).fit()
coef_A = modelo_A.params.to_dict()

# Incluye A y E para el modelo de mayores (consistente con  $dyE = \alpha * A - c_4 * E$ )
x_E = df[["Poblacion de adultos (15-64)", "Poblacion adultos mayores (65+)"]]
y_E = df["Cambio mayores"]
# x_E = sm.add_constant(x_E,prepend=False)
modelo_E = sm.OLS(
    y_E,
    x_E,

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).fit()
coef_E = modelo_E.params.to_dict()

print(modelo_J.summary())
print(modelo_A.summary())
print(modelo_E.summary())

```

#### OLS Regression Results

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=====
Dep. Variable:          Cambio juvenes   R-squared (uncentered):
0.824
Model:                  OLS              Adj. R-squared (uncentered):
0.804
Method:                Least Squares     F-statistic:
42.08
Date:                  Fri, 14 Nov 2025   Prob (F-statistic):
1.64e-07
Time:                  11:16:16          Log-Likelihood:
-244.04
No. Observations:      20               AIC:
492.1
Df Residuals:          18               BIC:
494.1
Df Model:              2
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t
[0.025      0.975]				
Poblacion de adultos (15-64)	-0.0085	0.001	-6.845	0.000
-0.011      -0.006				
Poblacion de juvenes (3-14)	0.0207	0.003	6.057	0.000
0.014      0.028				

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Omnibus:               13.709   Durbin-Watson:           1.332
Prob(Omnibus):         0.001   Jarque-Bera (JB):       13.941
Skew:                 -1.274   Prob(JB):               0.000939
Kurtosis:              6.199   Cond. No.               25.7
=====

```

#### Notes:

- [1]  $R^2$  is computed without centering (uncentered) since the model does not contain a constant.
- [2] Standard Errors assume that the covariance matrix of the errors is correctly specified.

# OLS Regression Results

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Dep. Variable:          Cambio adultos   R-squared (uncentered):
0.903
Model:                  OLS             Adj. R-squared (uncentered):
0.892
Method:                 Least Squares    F-statistic:
83.69
Date:                   Fri, 14 Nov 2025  Prob (F-statistic):
7.67e-10
Time:                   11:16:16         Log-Likelihood:
-286.99
No. Observations:      20              AIC:
578.0
Df Residuals:          18              BIC:
580.0
Df Model:               2
Covariance Type:       nonrobust
=====
=====

```

	coef	std err	t	P> t
[0.025      0.975]				
-----				
Poblacion de jovenes (3-14)	0.0359	0.029	1.225	0.236
-0.026      0.098				
Poblacion de adultos (15-64)	0.0036	0.011	0.340	0.738
-0.019      0.026				
=====				
Omnibus:	8.372	Durbin-Watson:		1.240
Prob(Omnibus):	0.015	Jarque-Bera (JB):		8.730
Skew:	-0.531	Prob(JB):		0.0127
Kurtosis:	6.057	Cond. No.		25.7
=====				

## Notes:

- [1]  $R^2$  is computed without centering (uncentered) since the model does not contain a constant.
- [2] Standard Errors assume that the covariance matrix of the errors is correctly specified.

# OLS Regression Results

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=====
Dep. Variable:          Cambio mayores   R-squared (uncentered):
0.903
Model:                  OLS             Adj. R-squared (uncentered):
0.893

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Method:                Least Squares    F-statistic:
84.09
Date:                  Fri, 14 Nov 2025    Prob (F-statistic):
7.38e-10
Time:                  11:16:16    Log-Likelihood:
-254.84
No. Observations:      20    AIC:
513.7
Df Residuals:          18    BIC:
515.7
Df Model:              2
Covariance Type:       nonrobust
=====
=====
                                coef    std err          t      P>|t|
-----
[0.025    0.975]
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Poblacion de adultos (15-64)    -0.0002     0.003    -0.076     0.940
-0.006     0.005
Poblacion adultos mayores (65+)  0.0344     0.026     1.325     0.202
-0.020     0.089
=====
Omnibus:                    13.437    Durbin-Watson:                1.473
Prob(Omnibus):              0.001    Jarque-Bera (JB):             19.057
Skew:                      0.937    Prob(JB):                     7.28e-05
Kurtosis:                   7.400    Cond. No.                     101.
=====

```

Notes:

- [1]  $R^2$  is computed without centering (uncentered) since the model does not contain a constant.
- [2] Standard Errors assume that the covariance matrix of the errors is correctly specified.

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[4]: # Graficacion

import numpy as np
import matplotlib.pyplot as plt

def sistema_edos(t, y):
    # y = [J, A, E]
    jt, at, et = y

    # COEFICIENTES
    c1 = -0.0085 # tasa de natalidad
    c2 = -0.0207 # tasa de mortalidad jóvenes

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#const1= -2.161e+06
tj=0.0359
c3 = -0.0036 # tasa de mortalidad adultos (positiva)
#const2=-3.146e+07
ta=-0.0002
c4 = -0.0344 # tasa de mortalidad mayores
#const3=-1.404e+06

# ECUACIONES
dyJ = c1*at-c2*jt#+const1
dyA = tj*jt-c3*at#+const2
dyE = ta*at-c4*et#+const3
return [dyJ, dyA, dyE]

def RK4(func, y0, t0, tf, h):
    t_values = np.arange(t0, tf + h, h)
    n = len(t_values)
    y_values = np.zeros((n, len(y0)), dtype=float)
    y_values[0] = y0

    for i in range(1, n):
        k1 = np.array(func(t_values[i-1], y_values[i-1]))
        k2 = np.array(func(t_values[i-1] + h/2, y_values[i-1] + h*k1/2))
        k3 = np.array(func(t_values[i-1] + h/2, y_values[i-1] + h*k2/2))
        k4 = np.array(func(t_values[i-1] + h, y_values[i-1] + h*k3))
        y_values[i] = y_values[i-1] + h*(k1 + 2*k2 + 2*k3 + k4)/6

    return t_values, y_values

# y0 = [J, A, E] en 2002
y0 = [27773497, 61886631, 5399307]
t0, tf, h = 0, 22, 0.01

t, y = RK4(sistema_edos, y0, t0, tf, h)

years = 2002 + t
fig, ax = plt.subplots(1, 1, figsize=(12, 8))
# Modelo
ax.plot(years, y[:, 0], 'b-', linewidth=2, label='Población de jóvenes')
ax.plot(years, y[:, 1], 'r-', linewidth=2, label='Población de adultos')
ax.plot(years, y[:, 2], 'g-', linewidth=2, label='Población de adultos mayores')
# Reales
years_real=df["ANIO"]
ax.plot(years_real,df["Poblacion de jovenes_
↪(3-14)"], 'bo', markersize=6, label="Real jovenes")
ax.plot(years_real,df["Poblacion de adultos_
↪(15-64)"], 'ro', markersize=6, label="Real adultos")

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ax.plot(years_real,df["Poblacion adultos mayores_↵
↵(65+)"], 'go',markersize=6,label="Real mayores")

ax.set_title('Población total por grupo')
ax.set_xlabel('Año')
ax.set_ylabel('Población')
ax.grid(True)
ax.legend()
plt.show()

```

