

modelacion_mexico

November 14, 2025

```
[11]: # Limpieza de datos

import pandas as pd

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

df=df.drop(['_id','RENGLON','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+)	...	Cambio adultos	Cambio mayores
0	2002	5399307	...	0.0	0.0
1	2003	5576695	...	1245096.0	177388.0
2	2004	5761262	...	1234986.0	184567.0
3	2005	5953231	...	1263623.0	191969.0
4	2006	6155725	...	1360846.0	202494.0
5	2007	6368873	...	1465313.0	213148.0
6	2008	6585485	...	1530449.0	216612.0
7	2009	6808607	...	1538706.0	223122.0
8	2010	7046284	...	1497157.0	237677.0
9	2011	7302545	...	1453767.0	256261.0
10	2012	7569447	...	1400902.0	266902.0
11	2013	7840039	...	1359450.0	270592.0

12	2014	8110040	...	1318887.0	270001.0
13	2015	8375584	...	1247533.0	265544.0
14	2017	8935919	...	2323830.0	560335.0
15	2018	9257526	...	1120796.0	321607.0
16	2019	9602115	...	1081726.0	344589.0
17	2020	9860625	...	942430.0	258510.0
18	2021	10027670	...	826442.0	167045.0
19	2022	10325860	...	924775.0	298190.0

[20 rows x 7 columns]

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[12]: df.to_csv('../data/mexico.csv')
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[24]: # 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 - c4*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,).fit()
coef_E=modelo_E.params.to_dict()

print(modelo_J.summary())
print(modelo_A.summary())
```

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print(modelo_E.summary())
```

OLS Regression Results

```
=====
Dep. Variable: Cambio jovenes 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 jovenes (3-14)	0.0207	0.003	6.057	0.000
0.014 0.028				
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	

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Notes:
[1] R2 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

```
=====
Dep. Variable: Cambio adultos R-squared (uncentered):
0.903
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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
 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]				
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.

[28]: # 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
    #const1= -2.161e+06
    tj=0.0359
    c3 = -0.0036 # tasa de mortalidad adultos (positiva)
    #const2=-3.146e+07
    ta=-0.0002
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c4 = 0.0344 # tasa de mortalidad mayores
#const3=-1.404e+06

# ECUACIONES
dyJ = c1*at-c2*tj#+const1
dyA = tj*tj-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\u2192(3-14)"],'bo',markersize=6,label="Real jovenes")
ax.plot(years_real,df["Poblacion de adultos\u2192(15-64)"],'ro',markersize=6,label="Real adultos")
ax.plot(years_real,df["Poblacion adultos mayores\u2192(65+)"],'go',markersize=6,label="Real mayores")

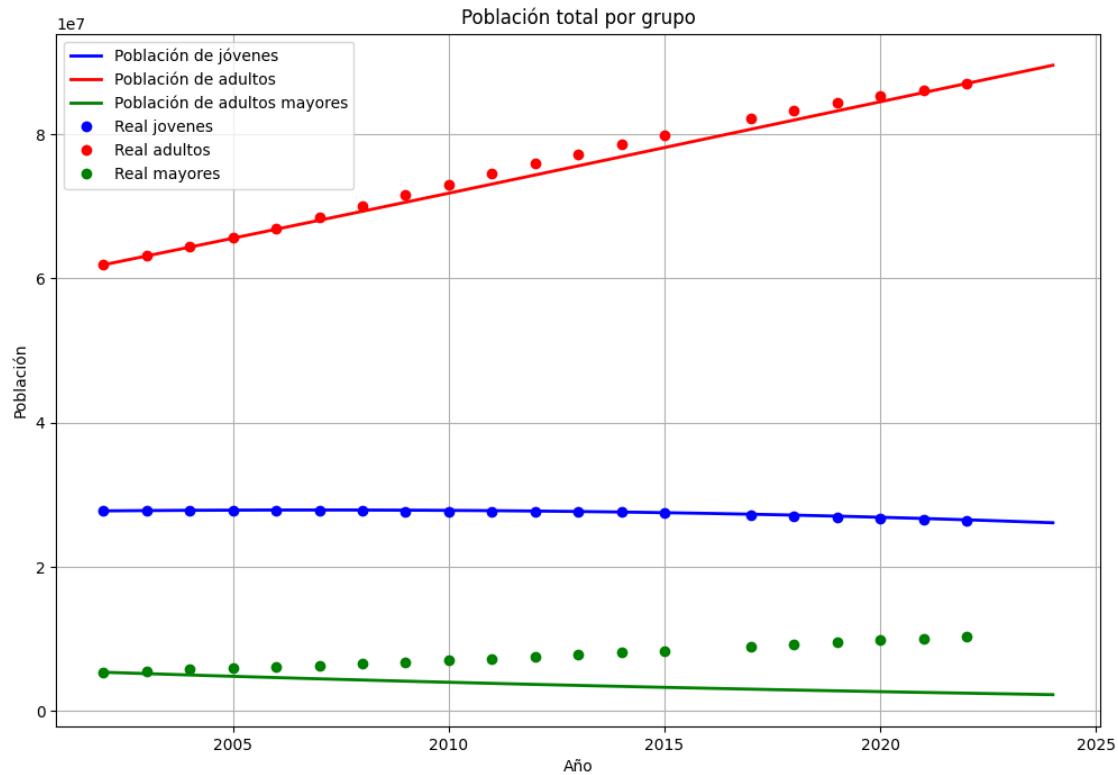
ax.set_title('Población total por grupo')
ax.set_xlabel('Año')

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ax.set_ylabel('Población')
ax.grid(True)
ax.legend()
plt.show()

```



```

[ ]: fig,ax = plt.subplots(1,1,figsize=(12,8))
ax.plot(df["Poblacion de adultos (15-64)"],linewidth=1,label="Poblacion de adultos")
ax.plot(df["Poblacion de jovenes (3-14)"],linewidth=1,label="Poblacion de jovenes")
ax.plot(df["Poblacion adultos mayores (65+)"],linewidth=1,label="Poblacion de adultos mayores")
ax.set_xlabel("Año")
ax.set_ylabel("Poblacion total")
ax.grid(True)
ax.legend()
plt.show

```

```
[ ]: <function matplotlib.pyplot.show(close=None, block=None)>
```

