

```

import pandas as pd
import numpy as np
import statsmodels.formula.api as smf
import matplotlib.pyplot as plt

def densityFunction(p, q, t):
    return (np.exp((p+q)*t)*p*(p+q)**2)/((p*np.exp((p+q)*t)+q)**2)

data = pd.read_csv('iphone_sales.txt', delim_whitespace=True)

def convert_quarter_to_date(quarter):
    q, y = quarter.split('_')
    q = int(q[1])
    y = int('20' + y)
    month = 3 * q
    if month > 12:
        month -= 12
        y += 1
    return pd.Timestamp(year=y, month=month, day=1)

data['date'] = data['Quarter'].apply(convert_quarter_to_date)

date = data['date']

sales = data['Sales_MM_units']

cum_sales = np.cumsum(sales)

cum_sales_squared = cum_sales**2

mod = smf.ols(formula='sales ~ cum_sales + cum_sales_squared', data=data)

res = mod.fit()

print(res.summary())

```



OLS Regression Results

```

=====
Dep. Variable:          sales    R-squared:                0.873
Model:                  OLS      Adj. R-squared:           0.865
Method:                 Least Squares    F-statistic:             109.9
Date:                   Sun, 07 Jul 2024    Prob (F-statistic):       4.61e-15
Time:                   19:54:01    Log-Likelihood:          -120.18
No. Observations:       35      AIC:                     246.4
Df Residuals:           32      BIC:                     251.0
Df Model:                2
Covariance Type:        nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
Intercept	3.6963	2.205	1.676	0.103	-0.795	8.188
cum_sales	0.1130	0.017	6.737	0.000	0.079	0.147
cum_sales_squared	-5.508e-05	2.11e-05	-2.610	0.014	-9.81e-05	-1.21e-05
=====						
Omnibus:		3.898	Durbin-Watson:		1.824	
Prob(Omnibus):		0.142	Jarque-Bera (JB):		2.617	
Skew:		0.627	Prob(JB):		0.270	
Kurtosis:		3.470	Cond. No.		4.32e+05	
=====						

Notes:

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

[2] The condition number is large, 4.32e+05. This might indicate that there are strong multicollinearity or other numerical problems.

```
b = res.params
b
```

```
Intercept      3.696307
cum_sales      0.112994
cum_sales_squared -0.000055
dtype: float64
```

```
m1 = (-b['cum_sales']+np.sqrt(b['cum_sales']**2 - 4*b['Intercept']*b['cum_sales_squared']))/(2*b['
m1
```

```
-32.206910098759515
```

```
m2 = (-b['cum_sales']-np.sqrt(b['cum_sales']**2 - 4*b['Intercept']*b['cum_sales_squared']))/(2*b['
m2
```

```
2083.8220174932867
```

```
m = max(m1, m2)
print(m)
```

```
2083.8220174932867
```

```
p = b['Intercept']/m
q = -m*b['cum_sales_squared']
print("p = ",p)
print("q = ", q)
```

```
p = 0.0017738112418998347
q = 0.11476751136366699
```

```
date_forecast = pd.date_range(start='2007-Q3', end='2030-Q3', freq='QS')
```

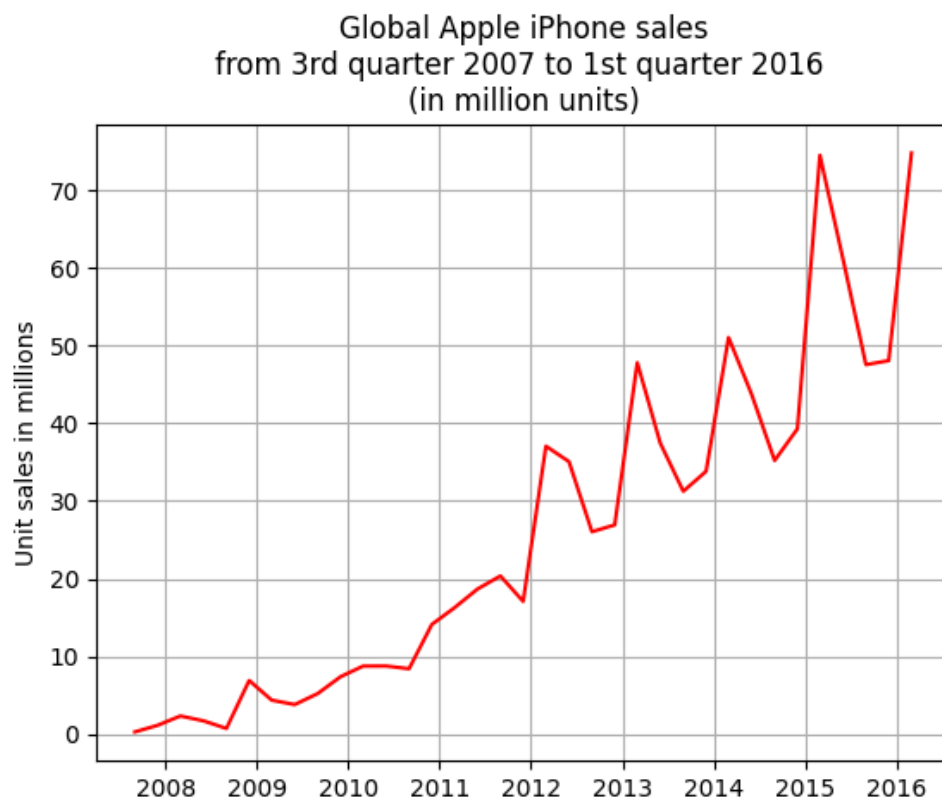
```
t = range(len(date_forecast))
```

```
f = densityFunction(p,q,t)
sales_forecast = m*f
```

```

fig, ax = plt.subplots()
ax.plot(date, sales, color='red')
plt.ylabel('Unit sales in millions')
#plt.xlabel('time')
plt.title("Global Apple iPhone sales\nfrom 3rd quarter 2007 to 1st quarter 2016 \n(in million unit
#plt.legend()
ax.grid(True)
plt.show()

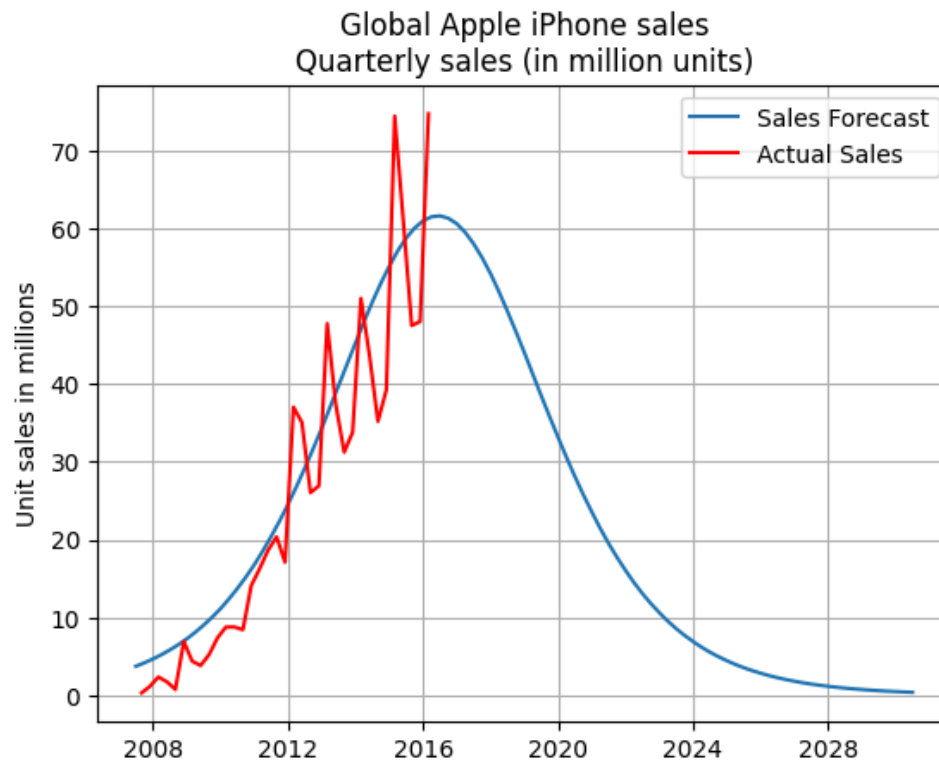
```



```

fig, ax = plt.subplots()
ax.plot(date_forecast, sales_forecast, label='Sales Forecast')
ax.plot(date, sales, color='red', label='Actual Sales')
plt.ylabel('Unit sales in millions')
#plt.xlabel('time')
plt.title("Global Apple iPhone sales\nQuarterly sales (in million units)")
plt.legend()
ax.grid(True)
plt.show()

```



```
sales_apple = sales
date_apple = date
date_forecast_apple = date_forecast
sales_forecast_apple = sales_forecast
```

✓ PEAK

```
peak_time = -1/(p+q)*np.log(p/q)
peak_time
```

```
35.77939457746893
```

```
print(data['Sales_MM_units'].idxmax() + 1)
```

```
35
```

✓ Samsung

```
data2 = pd.read_csv('galaxy_sales.csv')
```

```
data2.head()
```



	Quarter	sales	
0	Q1 '10	64.90	
1	Q2 '10	65.33	
2	Q3 '10	71.67	
3	Q4 '10	79.17	
4	Q1 '11	68.78	

Pasos siguientes:

Generar código con data2

Ver gráficos recomendados

```
quarter = data2['Quarter']
qs = quarter.str.replace(r"(Q\d) '(\d+)", r'20\2-\1', regex=True) # e.g., "Q1 '10" to "2010-Q1"
qs = qs.str.replace('*', '', regex=False) # Remove any asterisks if present


# Convert the cleaned string to a PeriodIndex and then to timestamps
data2['date'] = pd.PeriodIndex(qs.values, freq='Q').to_timestamp()

date = data2['date']

# Get coefficients
sales = data2['sales']

cum_sales = np.cumsum(sales)
cum_sales_squared = cum_sales**2

mod = smf.ols(formula='sales ~ cum_sales + cum_sales_squared', data=data2)
res = mod.fit()
print(res.summary())
```



OLS Regression Results											
=====											
Dep. Variable:	sales	R-squared:	0.821								
Model:	OLS	Adj. R-squared:	0.802								
Method:	Least Squares	F-statistic:	43.44								
Date:	Sun, 07 Jul 2024	Prob (F-statistic):	8.17e-08								
Time:	19:54:02	Log-Likelihood:	-73.418								
No. Observations:	22	AIC:	152.8								
Df Residuals:	19	BIC:	156.1								
Df Model:	2										
Covariance Type:	nonrobust										
=====											
	coef						std err	t	P> t	[0.025	0.975]

Intercept	53.7484	4.506	11.928	0.000	44.317	63.180					
cum_sales	0.0766	0.011	7.173	0.000	0.054	0.099					
cum_sales_squared	-2.806e-05	5.07e-06	-5.530	0.000	-3.87e-05	-1.74e-05					
=====											
Omnibus:	1.888	Durbin-Watson:	1.207								
Prob(Omnibus):	0.389	Jarque-Bera (JB):	1.056								
Skew:	0.116	Prob(JB):	0.590								
Kurtosis:	1.952	Cond. No.	5.30e+06								
=====											

Notes:

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

[2] The condition number is large, 5.3e+06. This might indicate that there are strong multicollinearity or other numerical problems.

```
b = res.params
b
```

```
↪ Intercept          53.748387
   cum_sales          0.076600
   cum_sales_squared  -0.000028
dtype: float64
```

```
m1 = (-b['cum_sales']+np.sqrt(b['cum_sales']**2 - 4*b['Intercept']*b['cum_sales_squared']))/(2*b['
m2 = (-b['cum_sales']-np.sqrt(b['cum_sales']**2 - 4*b['Intercept']*b['cum_sales_squared']))/(2*b['
```

```
m = max(m1, m2)
m
```

```
↪ 3308.96521821821
```

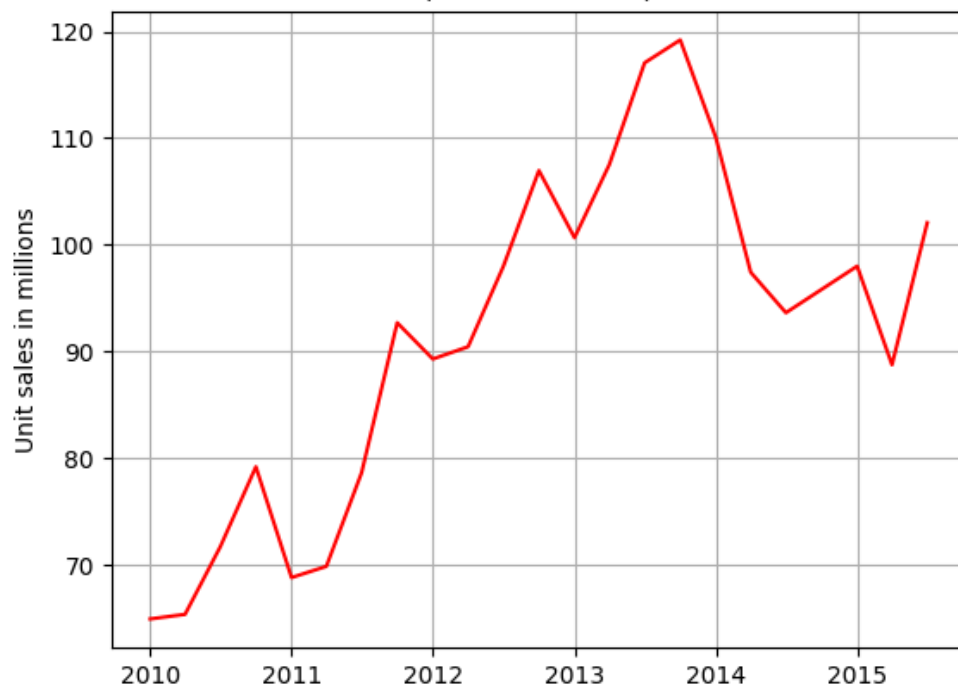
```
p = b['Intercept']/m
q = -m*b['cum_sales_squared']
print("p = ",p)
print("q = ", q)
```

```
↪ p = 0.016243261464994826
   q = 0.09284320017915797
```

```
fig, ax = plt.subplots()
ax.plot(date, sales, color='red')
plt.ylabel('Unit sales in millions')
#plt.xlabel('time')
plt.title("Global Samsung Galaxy sales\nfrom 1st quarter 2010 to 3rd quarter 2015 \n(in million ur
#plt.legend()
ax.grid(True)
plt.show()
```



Global Samsung Galaxy sales
from 1st quarter 2010 to 3rd quarter 2015
(in million units)



```
date_forecast = pd.date_range(start='2010-Q1', end='2030-Q3', freq='QS')
```

```
t = range(len(date_forecast))
f = densityFunction(p,q,t)
sales_forecast = m*f
```

```
fig, ax = plt.subplots()
ax.plot(date_forecast, sales_forecast, label='Sales Forecast')
ax.plot(date, sales, color='red', label='Actual Sales')
plt.ylabel('Unit sales in millions')
#plt.xlabel('time')
plt.title("Global Samsung Galaxy sales\nQuarterly sales (in million units)")
plt.legend()
ax.grid(True)
plt.show()
```



```
sales_samsung = sales
date_samsung = date
date_forecast_samsung = date_forecast
sales_forecast_samsung = sales_forecast
```



✓ Sales peak



```
peak_time = -1/(p+q)*np.log(p/q)
peak_time
```

```
15.98029563590453
```

```
print(data2['sales'].idxmax() + 1)
```

```
16
```

✓ Comparasion

```
fig, ax = plt.subplots()
ax.plot(date_forecast_samsung, sales_forecast_samsung, label='Sales Forecast Samsung')
ax.plot(date_samsung, sales_samsung, color='green', label='Actual Sales Samsung')
ax.plot(date_forecast_apple, sales_forecast_apple, label='Sales Forecast Apple')
ax.plot(date_apple, sales_apple, color='red', label='Actual Sales Apple')
plt.ylabel('Unit sales in millions')
#plt.xlabel('time')
plt.title("Global Apple iPhone and Samsung Galaxy sales\nQuarterly sales (in million units)")
plt.legend()
ax.grid(True)
plt.show()
```

