

HW-1

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0.0.1 Hripsime Soghomonyan - Bass Model Homework

1. 2024 innovation: <https://time.com/7094854/sakuu-kavian/>
2. I have chosen from 2024 innovations Sakuu's Kavian platform's 3D printing electrodes, and as a similar innovation from past graphite electrodes. Graphite electrodes are made by mixing graphite with binder and making it into solid blocks. This process requires heating at extremely high temperatures and toxic chemicals, which makes it harmful for the environment. 3D printed electrodes are made by layered printing technique, without using any liquid, which is faster, less energy consuming and more eco-friendly. Graphite electrodes are used in steel production, especially in electric arc and ladle furnaces, while 3D printed ones are used in lithium-ion battery production, which itself is used in the electric vehicle and other portable electronic devices' production.
3. Consumption volume of graphite electrodes worldwide 2014-2019: <https://www.statista.com/statistics/1235825/graphite-electrode-consumption-volume-worldwide/>

```
[2]: import numpy as np
import pandas as pd
import scipy.optimize as opt
import matplotlib.pyplot as plt
```

```
[3]: data = pd.read_excel('data/electrodes_data.xlsx')
print(data)
```

	Year	Volume
0	2014	1000.5
1	2015	949.5
2	2016	886.9
3	2017	856.4
4	2018	875.6
5	2019	892.0

```
[4]: years = data["Year"].values
volume = data["Volume"].values
```

```
[5]: #4
def bass_model(t, p, q, M):
```

```

    return (M * (p + q) ** 2 * np.exp(-(p + q) * t)) / (p + q * np.exp(-(p + q) * t)) ** 2

```

```
[6]: years -= years.min()
```

```
[7]: params, _ = opt.curve_fit(bass_model, years, volume, p0=[0.03, 0.38, 16000])
p, q, M = params
print(f"Estimated Parameters: p={p:.4f}, q={q:.4f}, M={M:.2f}")
```

Estimated Parameters: p=0.0134, q=-0.0135, M=970.80

```
[8]: # 5
def predict_diffusion(p, q, M, start_year=2020, end_year=2040):
    future_years = np.arange(start_year, end_year)
    adopters = np.zeros(len(future_years))
    for i in range(1, len(future_years)):
        adopters[i] = (p + q * (sum(adopters[:i]) / M)) * (M - sum(adopters[:i]))
    return future_years, np.cumsum(adopters)

graphite_years, graphite_adoption = predict_diffusion(p, q, M)

graphite_predictions_df = pd.DataFrame({"Year": graphite_years,
    "Predicted_Adopters": graphite_adoption})

print(graphite_predictions_df)
```

	Year	Predicted_Adopters
0	2020	0.000000
1	2021	13.004856
2	2022	25.662875
3	2023	37.987809
4	2024	49.992690
5	2025	61.689877
6	2026	73.091099
7	2027	84.207494
8	2028	95.049646
9	2029	105.627621
10	2030	115.950996
11	2031	126.028888
12	2032	135.869985
13	2033	145.482568
14	2034	154.874533
15	2035	164.053418
16	2036	173.026420
17	2037	181.800414
18	2038	190.381971
19	2039	198.777377

6. I am going to analyze the diffusion worldwide, as my data is about worldwide consumption, and not country specific. Country specific analysis is not compatible with the topic, as electrodes are mainly consumed during steel production, which is not country specific thing, therefore it needs global overview.

```
[10]: # 7
def estimate_adopters_by_period(p, q, M, start_year=2020, end_year=2040):
    future_years = np.arange(start_year, end_year)
    new_adopters = np.zeros(len(future_years))
    cumulative_adopters = np.zeros(len(future_years))

    for i in range(1, len(future_years)):
        new_adopters[i] = (p + q * (cumulative_adopters[i - 1] / M)) * (M -
        ↪cumulative_adopters[i - 1])
        cumulative_adopters[i] = cumulative_adopters[i - 1] + new_adopters[i]

    return future_years, new_adopters, cumulative_adopters

adoption_years, adopters_per_period, cumulative_adopters =
    ↪estimate_adopters_by_period(p, q, M)

adopters_df = pd.DataFrame({
    "Year": adoption_years,
    "New_Adopters": adopters_per_period,
    "Cumulative_Adopters": cumulative_adopters
})

print(adopters_df)
```

	Year	New_Adopters	Cumulative_Adopters
0	2020	0.000000	0.000000
1	2021	13.004856	13.004856
2	2022	12.658019	25.662875
3	2023	12.324934	37.987809
4	2024	12.004881	49.992690
5	2025	11.697187	61.689877
6	2026	11.401222	73.091099
7	2027	11.116395	84.207494
8	2028	10.842152	95.049646
9	2029	10.577975	105.627621
10	2030	10.323375	115.950996
11	2031	10.077893	126.028888
12	2032	9.841097	135.869985
13	2033	9.612582	145.482568
14	2034	9.391965	154.874533
15	2035	9.178885	164.053418
16	2036	8.973002	173.026420
17	2037	8.773994	181.800414

18	2038	8.581557	190.381971
19	2039	8.395405	198.777377