

# Curve Fitting (Mr. P Solver)

Video Link: <https://youtu.be/peBOquJ3fDo> (<https://youtu.be/peBOquJ3fDo>)

Codes: [https://www.youtube.com/redirect?](https://www.youtube.com/redirect?event=video_description&redir_token=QUFFLUhqa0lKYWJKTGs5amFORmhraS0tV1NCcjBIWDF1d3xBC)

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```
In [1]: import numpy as np
import matplotlib.pyplot as plt
import scipy as sp
import pandas as pd
from scipy.optimize import curve_fit
```

Given some  $x_{data} = [\dots]$  and  $y_{data} = [\dots]$  and a model function  $f$  that depends on unknown parameter(s)  $\beta$ . The goal is to find optimal set of parameters  $\beta$  such that the function  $y = f(x, \beta)$  best resembles the data.

1. One way of doing this is to minimize  $\sum_i (f(x_i, \beta) - y_i)^2$  by adjusting the parameters in  $\beta$ . Here  $x_i$  and  $y_i$  are the  $i^{th}$  data point. This is called **Least Square Method**.
2. If the  $y_i$  data points have corresponding errors on them, then minimizing  $\sum_i (f(x_i, \beta) - y_i)^2 / \sigma_i^2$  gives the **maximum likelihood** estimate for  $\beta$ .

Consider the model function  $f(x, \beta) = f(x, a, b, c) = a(x - b)^2 + c$  for a given data (where  $\beta = (a, b, c)$  are the optimal parameters). We fit the data in the following steps;

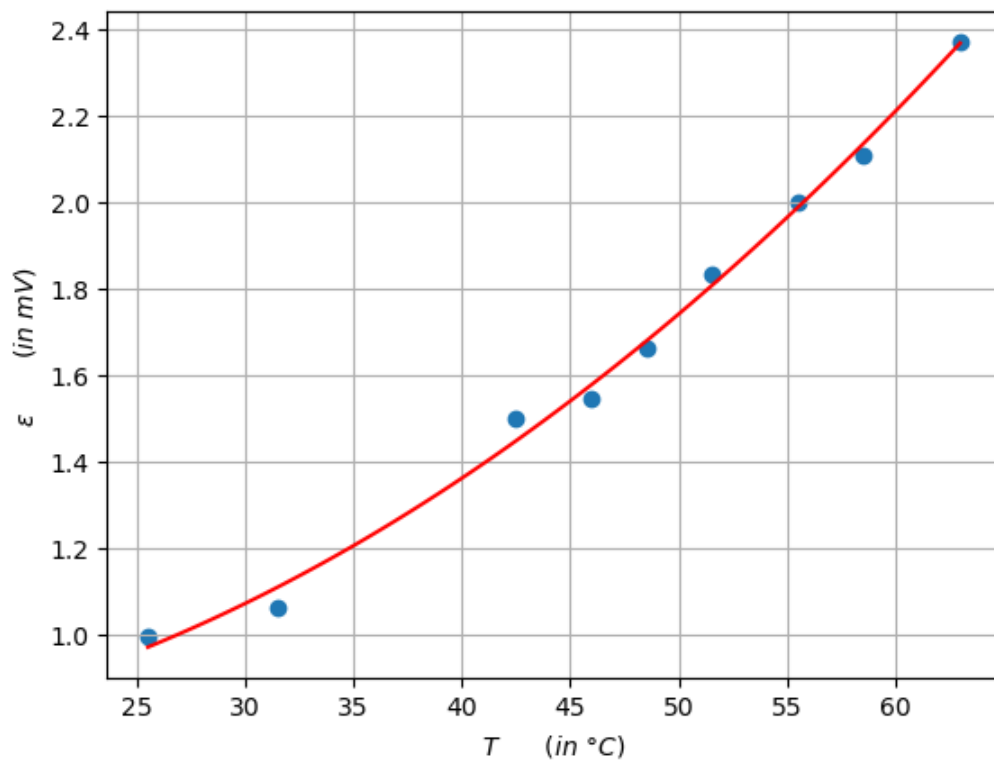
1. Define the model function.
2. Use `curve_fit` function of `scipy`. It requires initial guesses of  $\beta$ .

```
In [2]: x_data = np.array([25.5, 31.5, 42.5, 46, 48.5, 51.5, 55.5, 58.5, 63])
y_data = np.array([0.995, 1.06, 1.5, 1.545, 1.66, 1.833, 2, 2.108, 2.37])
```

```
In [3]: def model_f(x, a, b, c):
        return a*(x-b)**2 + c
```

```
In [4]: popt, pcov = curve_fit(model_f, x_data, y_data, p0=[3, 2, -16])
a_opt, b_opt, c_opt = popt
x_model = np.linspace(min(x_data), max(x_data), 100)
y_model = model_f(x_model, a_opt, b_opt, c_opt)
```

```
In [5]: plt.scatter(x_data,y_data)
plt.plot(x_model,y_model, color='r')
plt.xlabel('$T$ \t $(in$ $\text{degree C})$')
plt.ylabel('$\epsilon$ \t $(in$ $mV)$')
plt.grid()
plt.show()
```



curve\_fit gives *optimal parameters* popt and the corresponding *covariances* pcov .

```
In [ ]:
```