

PT-100 Lab Assignment

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Abstract—This document contains a lab report on the modeling of the voltage-temperature characteristics of the PT-100 RTD (Resistance Temperature Detector) using least squares method.

1 TRAINING DATA

The training data gathered by the PT-100 to train the Arduino is shown in Table 1.

Temperature (°C)	Voltage (V)
66	1.85
27	1.76
2	1.66
23	1.72
56	1.82
34	1.76
33	1.75
31	1.74

TABLE 1: Training data.

The C++ source codes/data.cpp was used along with *platformio* to drive the Arduino. The effective schematic circuit diagram is shown in Figure 1.

2 MODEL

For the PT-100, we use the Callendar-Van Dusen equation

$$V(T) = V(0) \left(1 + AT + BT^2 \right) \quad (1)$$

$$\Rightarrow c = \mathbf{n}^T \mathbf{x} \quad (2)$$

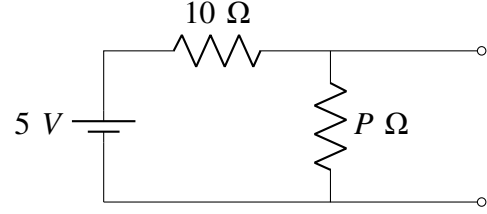


Fig. 1: Schematic Circuit Diagram to Measure the Output of PT-100 (P).

where

$$c = V(T), \quad \mathbf{n} = V(0) \begin{pmatrix} 1 \\ A \\ B \end{pmatrix}, \quad \mathbf{x} = \begin{pmatrix} 1 \\ T \\ T^2 \end{pmatrix} \quad (3)$$

For multiple points, (2) becomes

$$\mathbf{X}^T \mathbf{n} = \mathbf{C} \quad (4)$$

where

$$\mathbf{X} = \begin{pmatrix} 1 & 1 & \dots & 1 \\ T_1 & T_2 & \dots & T_n \\ T_1^2 & T_2^2 & \dots & T_n^2 \end{pmatrix} \quad (5)$$

$$\mathbf{C} = \begin{pmatrix} V(T_1) \\ V(T_2) \\ \vdots \\ V(T_n) \end{pmatrix} \quad (6)$$

and \mathbf{n} is the unknown.

3 SOLUTION

We approximate \mathbf{n} by using the least squares method. The Python code codes/lsq.py solves for \mathbf{n} .

The calculated value of \mathbf{n} is

$$\mathbf{n} = \begin{pmatrix} 1.6547 \\ 3.199 \times 10^{-3} \\ -3.9599 \times 10^{-6} \end{pmatrix} \quad (7)$$

The approximation is shown in Fig. 2.

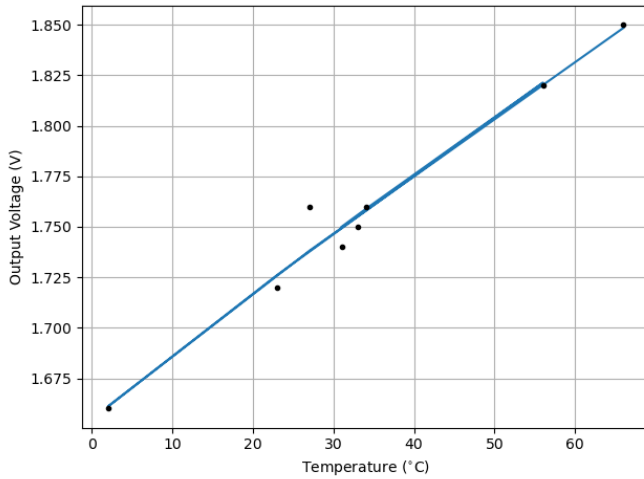


Fig. 2: Training the model.

Thus, the approximate model is given by

$$V(T) = 1.6547 + (3.199 \times 10^{-3})T - (3.9599 \times 10^{-6})T^2 \quad (8)$$

Notice in (8) that the coefficient of T^2 is negative, and hence the governing function is strictly concave. Hence, we cannot use gradient descent methods to solve this problem.

4 VALIDATION

The validation dataset is shown in Table 2. The results of the validation are shown in Fig. 3.

Temperature (°C)	Voltage (V)
4	1.67
25	1.73
61	1.83
35	1.77

TABLE 2: Validation data.

5 CONCLUSION

This lab experiment demonstrates how machine learning methods can be used to model the behaviour of an unknown device, and find the right parameters that fit the model. It also shows how to use Python libraries and frameworks to collect data and perform optimization.

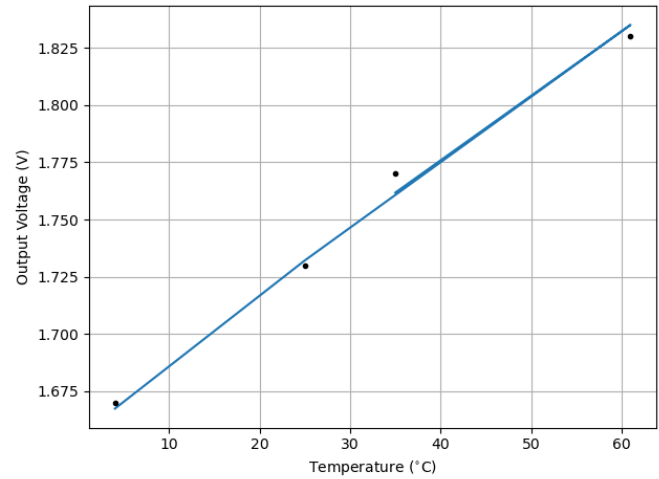


Fig. 3: Validating the model.