

PyLab - Ohm and Power laws

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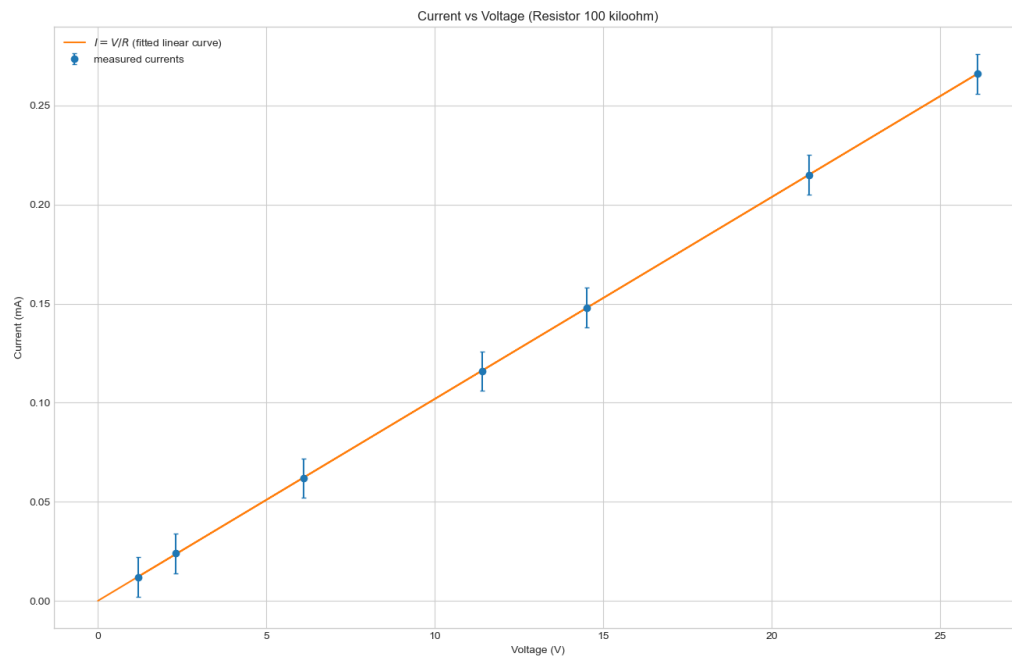
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Exercise 1: Introduction to fitting methods

1 Procedures

Voltage (V)	Current (mA)	Resistance (kOhm)
1.2	0.012	100
2.3	0.024	95.83333333
6.1	0.062	98.38709677
11.4	0.116	98.27586207
14.5	0.148	97.97297297
21.1	0.215	98.13953488
26.1	0.266	98.12030075

2 Analysis



3 Conclusions

Exercise 3: Nonlinear fitting methods II

Appendix

Python Code: Exercise 1

```
import numpy as np
import scipy.optimize as optim
import matplotlib.pyplot as plt

def chi2(y_measure, y_predict, errors):
    """Calculate the chi squared value given a measurement with errors and
    prediction"""
    return np.sum( np.power(y_measure - y_predict, 2) / np.power(errors, 2) )

def chi2reduced(y_measure, y_predict, errors, number_of_parameters):
    """Calculate the reduced chi squared value given a measurement with error
    and prediction,
    and knowing the number of parameters in the model."""
    return chi2(y_measure, y_predict, errors)/(y_measure.size - number_of_par

# we have constant voltage uncertainty which is 0.1 V
voltage_uncertainty = 0.1

def current_uncertainty(current):
    """return the uncertainty in current for given values of current"""
    if current > 100:
        return 1
    elif current > 10:
        return 0.1
    else:
        return 0.01

# model function
def linear_model_function(x, a, b):
    return a*x + b

# filename
filename = "100k.csv"

# load the csv file as txt
measured_voltages, measured_currents = np.loadtxt(filename,
                                                    skiprows=1,
                                                    usecols=(0,1),
                                                    delimiter=",",
```

```

unpack=True)

# create error array for the voltage
voltage_errors = np.ones_like(measured_voltages) * voltage_uncertainty

# create error array for the current
current_errors = np.vectorize(current_uncertainty)(measured_currents)

# do the curve fitting
popt, pcov = optim.curve_fit(linear_model_function,
                             measured_voltages,
                             measured_currents,
                             absolute_sigma=True,
                             sigma=current_errors)

print("linear_fit_model_gives_a=%.2f, b=%.2f" % (popt[0], popt[1]))
pvar = np.diag(pcov)

# new figure for this file
plt.figure(figsize=(16, 10))
plt.style.use("seaborn-whitegrid")

# plot the error bar chart
plt.errorbar(measured_voltages,
             measured_currents,
             yerr=current_errors,
             marker="o",
             label="measured_currents",
             capsize=2,
             ls="")

# plot the fitted curve
# add 0 to the measured data set
measured_voltages_with_0 = np.append(measured_voltages, 0)
plt.plot(measured_voltages_with_0,
         linear_model_function(measured_voltages_with_0, popt[0], popt[1]),
         label='I = V/R$(fitted_linear_curve)')

# legend and title
plt.title("Current_vs_Voltage_(Resistor_100_kiloohm)")
plt.xlabel("Voltage_(V)")
plt.ylabel("Current_(mA)")
plt.legend(loc="upper_left")
plt.savefig("lab_1_ex_1_plot.png")

```

```

chi2r = chi2reduced(measured_currents ,
                    linear_model_function(measured_voltages , popt[0] , popt[1]
                    current_errors ,
                    1)

print("fitted (average) resistance = %.3f_kiloohm" % (1/popt[0]))
print("error in fitted resistance = %.5f_kiloohm" % np.sqrt(pvar[0]))
print("model chi2r = %.3f" % chi2r)

```

Python Code: Exercise 3

```
import numpy as np
import scipy.optimize as optim
import matplotlib.pyplot as plt

def chi2(y_measure, y_predict, errors):
    """Calculate the chi squared value given a measurement with errors and
    prediction"""
    return np.sum( np.power(y_measure - y_predict, 2) / np.power(errors, 2) )

def chi2reduced(y_measure, y_predict, errors, number_of_parameters):
    """Calculate the reduced chi squared value given a measurement with error
    and prediction,
    and knowing the number of parameters in the model."""
    return chi2(y_measure, y_predict, errors)/(y_measure.size - number_of_parameters)

# we have constant voltage uncertainty which is 0.1 V
voltage_uncertainty = 0.1

def current_uncertainty(current):
    """return the uncertainty in current for given values of current"""
    if current > 100:
        return 1
    elif current > 10:
        return 0.1
    else:
        return 0.01

# model function
def linear_model_function(x, a, b):
    return a*x + b

# filename
filename = "100k.csv"

# load the csv file as txt
measured_voltages, measured_currents = np.loadtxt(filename,
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# create error array for the current
current_errors = np.vectorize(current_uncertainty)(measured_currents)

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popt, pcov = optim.curve_fit(linear_model_function,
                             measured_voltages,
                             measured_currents,
                             absolute_sigma=True,
                             sigma=current_errors)

print("linear_fit_model_gives_a=%.2f, b=%.2f" % (popt[0], popt[1]))
pvar = np.diag(pcov)

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plt.figure(figsize=(16, 10))
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# plot the fitted curve
# add 0 to the measured data set
measured_voltages_with_0 = np.append(measured_voltages, 0)
plt.plot(measured_voltages_with_0,
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plt.title("Current_vs_Voltage_(Resistor_100_kiloohm)")
plt.xlabel("Voltage_(V)")
plt.ylabel("Current_(mA)")
plt.legend(loc="upper_left")
plt.savefig("lab_1_ex_1_plot.png")

chi2r = chi2reduced(measured_currents,
                    linear_model_function(measured_voltages, popt[0], popt[1])
                    current_errors,

```


1)

```
print("fitted (average) resistance = %.3f kilohm" % (1/popt[0]))  
print("error in fitted resistance = %.5f kilohm" % np.sqrt(pvar[0]))  
print("model chi2r = %.3f" % chi2r)
```

References

[1]