

Investigation Into The Effect of Polarizers and Waveplates on a Helium-Neon Laser

Aditya K. Rao^{a)}

University of Toronto

MP222, 60 St. George Street, Toronto, Ontario M5S 1A7, Canada.

^{a)}adi.rao@mail.utoronto.ca; Student Number: 1008307761

Abstract. ABSTRACT

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INTRODUCTION

Electromagnetic waves are modeled as transverse plane waves that oscillate in the direction perpendicular to their propagation. The electric field of these waves can be described by (1).

$$\vec{E}_0 = E_i \hat{i} + E_j \hat{j} \quad (1)$$

One properties of such waves is their ability to become 'polarized'. This annihilates all other oscillations apart from a single plane. This can be achieved by passing the wave through a polarizer. The intensity of the light that passes through the polarizer is given by (2).

$$I = I_0 \cos^2(\theta) \quad (2)$$

In this experiment, the power is recorded as opposed to the intensity. However, because the power is proportional to the intensity, the same equation can be used. The power of an Electromagnetic wave is directly proportional to the product of its electric field and its complex conjugate. This is given by (3).

$$P \propto |E_0|^2 \quad (3)$$

Waveplates are another common optical element which, instead of restricting the oscillations of the electric field, shift the phase of the wave. This is done by changing the refractive index of the material.

$$\hat{\Pi}_\theta = \begin{bmatrix} \cos^2(\theta) & \sin(\theta)\cos(\theta) \\ \sin(\theta)\cos(\theta) & \sin^2(\theta) \end{bmatrix} \quad (4)$$

$$\hat{J} = \chi \hat{\Pi}_\theta + \zeta \hat{\Pi}_\theta' \quad (5)$$

Here, two different types of waveplate will be investigated: quarter and half waveplates. The quarter waveplate shifts the phase of the wave by $\pi/2$ radians. The half waveplate shifts the phase of the wave by π radians.

The resultant effect on the power should be that the power is unchanged for the half wave plate and that the power is reduced for the quarter waveplate due to destructive and constructive interference of the E_x and E_y components.

The jones matrix for a waveplate is given by (5). Here, χ and ζ are the complex amplitudes of the electric field. For the quarter waveplate, the Jones matrix is given by (6). For the half waveplate, the Jones matrix is given by (7).

$$\hat{J}_{QWP} = \begin{bmatrix} 1 & 0 \\ 0 & \exp\left\{\frac{i\pi}{2}\right\} \end{bmatrix} \quad (6) \quad \hat{J}_{HWP} = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} \quad (7)$$

METHODOLOGY

Two experiments were conducted in order to test and verify the effects of polarizers and waveplates on a Helium-Neon (HeNe) laser with an additional third being planned. The experimental setup for each is shown in Figures 1 and 2 respectively.

For each experiment, a PDA015C2 photodiode was used to measure the intensity of the light. The photodiode was connected to a DSO oscilloscope which was used to measure the voltage output from the optical setup. The HeNe laser was the first component attached to the optical breadboard. All other components were then roughly placed in position (not clamped) after which fine & coarse adjustments were made to align the beam path.

To reduce ambient light a shroud was put over the photodiode. The resultant baseline voltage dropped from 2V to $\approx 240 \pm 10$ mV.

Polarizers

The second polarizer was removed from the post holder. The polarizer's angle was then adjusted such that the outputted power delivered to the photodiode was below its saturation voltage. This maximum power output was recorded at $V = 5.00$ V. The angle of the polarizer was recorded as $157.69^\circ \pm 0.01^\circ$.

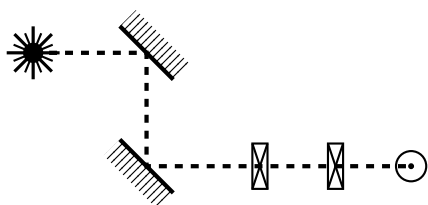


FIGURE 1: Experimental Setup for Experiment 1

This was kept the same for subsequent measurements. The second polarizer was then varied in angle from 0° to 270° in increments of 10° . The voltage output from the photodiode was recorded for each angle.

To reduce ambient light a shroud was put over the photodiode. The resultant baseline voltage dropped from 2V to $\approx 240 \pm 10$ mV.

Upon adding the second polarizer, massive fluctuations in the output voltage were observed (on the order of at least 0.5V, but proportional to the output voltage). This was likely due to an issue with the battery of the photodiode. However, upon replacing the battery, no change in the output was observed. Therefore, the issue was likely within the output of the HeNe laser. This is documented further later in the report. In order to mitigate this abnormality, the output voltage was recorded at specific time intervals where the output voltage was semi-stable.

Waveplates

The experimental setup in Fig. 1 was modified to place a waveplate between the two polarizers. In the first part of the experiment, the effect of a quarter waveplate was tested. The waveplate was placed at an angle of 0° with respect to the horizontal. The second polarizer was then varied in angle from 0° to 220° in increments of 20° . The voltage output from the photodiode was recorded for each angle.

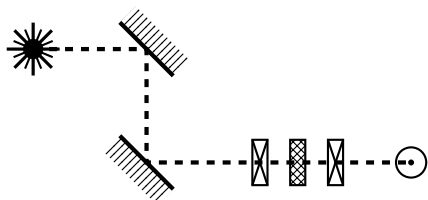


FIGURE 2: Experimental Setup for Experiment 2

This was then followed by a similar experiment with a half waveplate. The waveplate was placed at an angle of 0° with respect to the horizontal. The second polarizer was then varied in angle from 0° to 220° in increments of 20° . The voltage output from the photodiode was recorded for each angle.

Dielectrics

Due to time constraints, this experiment was not completed in full. The proposed setup is shown in Fig. 3. The setup was to be similar to the previous two experiments, with the addition of a dielectric material between the two polarizers. The dielectric material was to be placed at an angle of 0° with respect to the horizontal. The second polarizer was then varied in angle from 0° to 220° in increments of 20° . The voltage output from the photodiode was to be recorded for each angle.

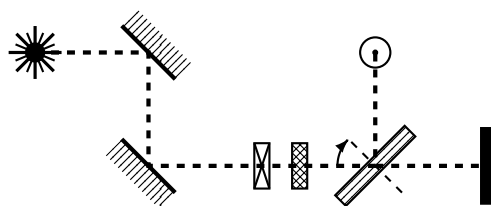


FIGURE 3: Proposed Experimental Setup for Experiment 3

The proposed experimental procedure would have been similar to the previous two experiments. However, instead of varying the angle of the polarizer, instead, the dielectric would have been rotated through 180° in increments of 10° . The resultant voltage output would have been recorded for each angle.

RESULTS & ANALYSIS

Quarter Waveplate

Polarizers

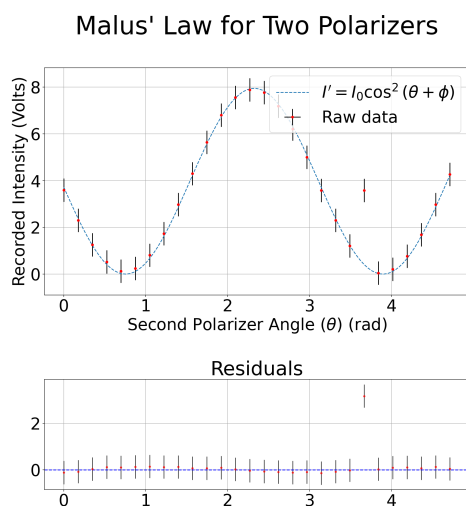


FIGURE 4: Voltage Output vs. Polarizer Angle. $\chi^2_{red} = 1.58$, $I_0 = 8.0 \pm 0.2$ V, $\phi = 0.82 \pm 0.02$

It can clearly be seen in Fig. 4 that the data adheres well to theory. All of the data points, apart from one outlier, falls well within the estimated uncertainty of the fitted curve given in (2). It should be stated that the χ^2 value of 1.58 is slightly high, however, the lack of a distinct pattern in the residuals indicate that this is a good fit regardless.

| Parameter | Value | Uncertainty |
|-----------|-------|-------------|
| I_0 | 8.0 | 0.2 |
| ϕ | 0.82 | 0.02 |

Recorded Intensity for Quarter Waveplate

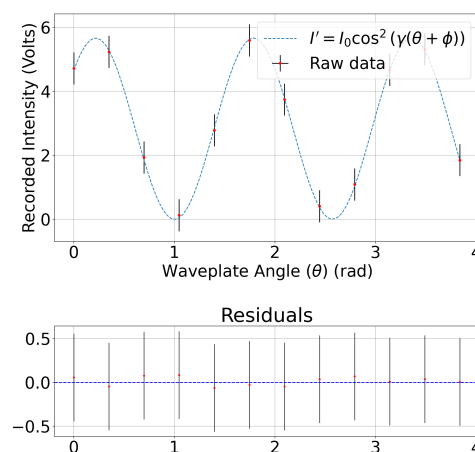


FIGURE 5: Voltage Output vs. Quarter Waveplate Angle. $\chi^2_{red} = 0.01$, $I_0 = 5.66 \pm 0.03$ V, $\phi = -0.218 \pm 0.004$, $\gamma = 2.001 \pm 0.003$

In Fig. 5, the data adheres well to the theoretical curve. The χ^2 value of 0.01 indicates that the data is a very good fit to the theoretical curve. The residuals are also randomly distributed around zero, indicating that the fit is good.

| Parameter | Value | Uncertainty |
|-----------|---------|-------------|
| I_0 | 5.66 | 0.03 |
| ϕ | -0.2175 | 0.004 |
| γ | 2.001 | 0.003 |

Half Waveplate

Recorded Intensity for Half Waveplate

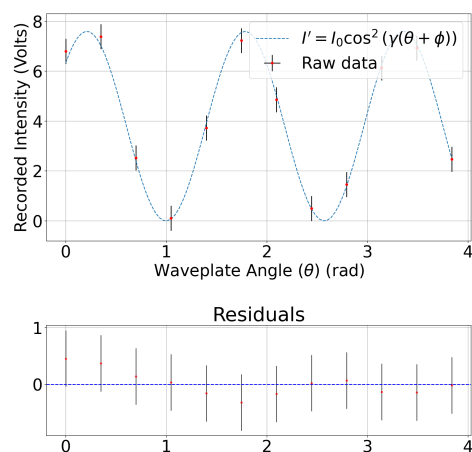


FIGURE 6: Voltage Output vs. Half Waveplate Angle. $\chi^2_{red} = 0.25$, $I_0 = 7.6 \pm 0.1$ V, $\phi = -0.21 \pm 0.01$, $\gamma = 2.00 \pm 0.01$

| Parameter | Value | Uncertainty |
|-----------|-------|-------------|
| I_0 | 7.6 | 0.1 |
| ϕ | -0.21 | 0.01 |
| γ | 2.00 | 0.01 |

DISCUSSION

An interesting result can be observed in comparing Fig. 6 and Fig. 4. The recorded maximum intensity I_0 is almost (bar a factor $V = 0.4$ V) identical despite the polarizers being set at angles to pass the minimum amount of light. This implies that at $\theta = \frac{\pi}{2}$, the half waveplate effectively makes the second polarizer transparent.

CONCLUSION

ACKNOWLEDGMENTS

The work conducted by the other lab partners was instrumental in this lab. Thank you to Jack and Billy for their help in setting up the equipment and conducting the experiments. Additionally, thank you to the Teaching Assistant Michael Sloan and Professor Boris Braverman for their guidance and support.

REFERENCES

1. B. Braverman, "Lab 1: Polarization," (2025).
2. G. Van Rossum and F. L. Drake, *Python 3 Reference Manual* (CreateSpace, Scotts Valley, CA, 2009).
3. C. R. Harris, K. J. Millman, S. J. van der Walt, R. Gommers, P. Virtanen, D. Cournapeau, E. Wieser, J. Taylor, S. Berg, N. J. Smith, R. Kern, M. Picus, S. Hoyer, M. H. van Kerkwijk, M. Brett, A. Haldane, J. F. del Río, M. Wiebe, P. Peterson, P. Gérard-Marchant, K. Sheppard, T. Reddy, W. Weckesser, H. Abbasi, C. Gohlke, and T. E. Oliphant, "Array programming with NumPy," *Nature* **585**, 357–362 (2020).
4. The pandas development team, "pandas-dev/pandas: Pandas," (2020).
5. J. Doe, private communication (2024), assistance Received.

Appendix

Raw Data

Experiment 1: Polarizers

| Polarizer #1 (°) | Polarizer #2 (°) | Voltage (V) |
|--------------------|--------------------|-------------|
| 159.69 | NaN | 0.05 |
| 159.69 | 0 | 3.6 |
| 159.69 | 10 | 2.3 |
| 159.69 | 20 | 1.26 |
| 159.69 | 30 | 0.53 |
| 159.69 | 40 | 0.13 |
| 159.69 | 50 | 0.24 |
| 159.69 | 60 | 0.81 |
| 159.69 | 70 | 1.74 |
| 159.69 | 80 | 2.98 |
| 159.69 | 90 | 4.3 |
| 159.69 | 100 | 5.64 |
| 159.69 | 110 | 6.81 |
| 159.69 | 120 | 7.56 |
| 159.69 | 130 | 7.89 |
| 159.69 | 140 | 7.77 |
| 159.69 | 150 | 7.19 |
| 159.69 | 160 | 6.21 |
| 159.69 | 170 | 5 |
| 159.69 | 180 | 3.58 |
| 159.69 | 190 | 2.3 |
| 159.69 | 200 | 1.21 |
| 159.69 | 210 | 3.58 |
| 159.69 | 220 | 0.05 |
| 159.69 | 230 | 0.21 |
| 159.69 | 240 | 0.77 |
| 159.69 | 250 | 1.69 |
| 159.69 | 260 | 2.98 |
| 159.69 | 270 | 4.27 |

Experiment 2: Quarter Waveplate

| Polarizer #1 (°) | Polarizer #2 (°) | QWP (°) | Voltage (V) |
|--------------------|--------------------|-----------|-------------|
| 159.69 | 220 | 0 | 4.71 |
| 159.69 | 220 | 20 | 5.23 |
| 159.69 | 220 | 40 | 1.93 |
| 159.69 | 220 | 60 | 0.13 |
| 159.69 | 220 | 80 | 2.78 |
| 159.69 | 220 | 100 | 5.59 |
| 159.69 | 220 | 120 | 3.74 |
| 159.69 | 220 | 140 | 0.41 |
| 159.69 | 220 | 160 | 1.09 |
| 159.69 | 220 | 180 | 4.67 |
| 159.69 | 220 | 200 | 5.31 |
| 159.69 | 220 | 220 | 1.85 |

Experiment 2: Half Waveplate

| Polarizer #1 (°) | Polarizer #2 (°) | HWP (°) | Voltage (V) |
|--------------------|--------------------|-----------|-------------|
| 159.69 | 220 | 0 | 6.8 |
| 159.69 | 220 | 20 | 7.39 |
| 159.69 | 220 | 40 | 2.53 |
| 159.69 | 220 | 60 | 0.11 |
| 159.69 | 220 | 80 | 3.73 |
| 159.69 | 220 | 100 | 7.23 |
| 159.69 | 220 | 120 | 4.86 |
| 159.69 | 220 | 140 | 0.5 |
| 159.69 | 220 | 160 | 1.46 |
| 159.69 | 220 | 180 | 6.13 |
| 159.69 | 220 | 200 | 6.93 |
| 159.69 | 220 | 220 | 2.47 |

Analysis Code

Experiment 1: Polarizers

```

1  #!/usr/bin/env python3
2  # -*- coding: utf-8 -*-
3  """
4  Created on Mon Jan 27 11:35:21 2025
5
6  @author: Aditya K. Rao
7  @github: @adirao-projects
8  """
9

```



```
10 import numpy as np
11 import pandas as pd
12 import toolkit as tk
13 import matplotlib.pyplot as plt
14
15 # uncert in voltage +- 0.01 V
16
17 def load_data(file):
18     df = pd.read_csv(file)
19     df = df.dropna()
20
21     # Uncertainty in Output Voltage
22     w_uncert = np.full(df.shape[0]+1, 0.5)
23
24     # Uncertainty in Angle
25     th_uncert = np.full(df.shape[0]+1, 0.01)
26
27     df['Wu'] = pd.Series(w_uncert)
28     df['Tu'] = pd.Series(th_uncert)
29
30     return df
31
32
33 def model_func_malus(theta, I0, phi):
34     return I0*((np.cos(theta+phi))**2)
35
36 def deg_rad(angle):
37     return (angle/180)*(np.pi)
38
39 def rad_deg(angle):
40     return (angle/np.pi)*(180)
41
42
43 def fit_plot(df):
44
45     xdata = deg_rad(df['Pb'].to_numpy())
46     ydata = df['W'].to_numpy()
47     y_unc = df['Wu'].to_numpy()
48     x_unc = deg_rad(df['Tu'].to_numpy())
49
50     #plt.errorbar(xdata, ydata, yerr=y_unc, xerr=x_unc, fmt='o')
51
52     #print(xdata)
53
54     data = tk.curve_fit_data(xdata, ydata, fit_type='custom',
55                             model_function_custom=model_func_malus,
56                             uncertainty=y_unc, uncertainty_x=x_unc,
57                             res=True, chi=True, guess=(7, np.pi/4))
58
59
60     meta = {'title' : "Malus' Law for Two Polarizers\n",
61            'xlabel' : r'Second Polarizer Angle ($\theta$) (rad)',
62            'ylabel' : 'Recorded Intensity (Volts)',
63            'chisq' : data['chisq'],
```

```

64         'fit-label': r"$I' = I_0 \cos^2 (\theta+\phi)$",
65         'data-label': "Raw data",
66         'save-name' : 'Malus',
67         'loc' : 'upper right'}
68
69     tk.quick_plot_residuals(xdata, ydata, data['plotx'], data['ploty'],
70                             data['residuals'], meta=meta,
71                             uncertainty=y_unc, uncertainty_x=x_unc,
72                             save=True)
73
74     return data['chisq'], data['popt'], data['pstd']
75
76 if __name__ == '__main__':
77     df = load_data('part1.csv')
78
79     params = fit_plot(df)
80
81     printvals = [r'$\chi_{red}^2+f' = {params[0]}$']
82     for i,v in enumerate([r'I_0', r'\phi']):
83         printvals.append(f'$v = {params[1][i]} \pm {params[2][i]}$')
84
85     tk.block_print(printvals, 'Fit Parameters for Two Polarizer')

```

Experiment 2: Waveplates

```

1  #!/usr/bin/env python3
2  # -*- coding: utf-8 -*-
3  """
4  Created on Mon Feb 10 09:33:31 2025
5
6  @author: Aditya K. Rao
7  @github: @adiraio-projects
8  """
9
10 import numpy as np
11 import pandas as pd
12 import toolkit as tk
13 import matplotlib.pyplot as plt
14
15 # uncert in voltage +- 0.01 V
16
17 def load_data(file):
18     df = pd.read_csv(file)
19     df = df.dropna()
20
21     # Uncertainty in Output Voltage
22     w_uncert = np.full(df.shape[0]+1, 0.5)
23
24     # Uncertainty in Angle
25     th_uncert = np.full(df.shape[0]+1, 0.01)
26

```

```

27     df['Wu'] = pd.Series(w_uncert)
28     df['Tu'] = pd.Series(th_uncert)
29
30     return df
31
32
33 def model_func_wp(theta, I0, phi, phase):
34     return I0*((np.cos(phase*(theta+phi)))**2)
35
36 def deg_rad(angle):
37     return (angle/180)*(np.pi)
38
39 def rad_deg(angle):
40     return (angle/np.pi)*(180)
41
42
43 def fit_plot(df, plate):
44
45     xdata = deg_rad(df[plate].to_numpy())
46     ydata = df['W'].to_numpy()
47     y_unc = df['Wu'].to_numpy()
48     x_unc = deg_rad(df['Tu'].to_numpy())
49
50     #plt.errorbar(xdata, ydata, yerr=y_unc, xerr=x_unc, fmt='o')
51
52     #print(xdata)
53
54     data = tk.curve_fit_data(xdata, ydata, fit_type='custom',
55                             model_function_custom=model_func_wp,
56                             uncertainty=y_unc, uncertainty_x=x_unc,
57                             res=True, chi=True, guess=(8, 0.5, 1.5))
58
59     if plate == 'QWP':
60         plate_name = 'Quarter'
61
62     elif plate == 'HWP':
63         plate_name = 'Half'
64
65     meta = {'title' : f"Recorded Intensity for {plate_name} Waveplate\n",
66            'xlabel' : r'Waveplate Angle ( $\theta$ ) (rad)',
67            'ylabel' : 'Recorded Intensity (Volts)',
68            'chisq' : data['chisq'],
69            'fit-label': r" $I = I_0 \cos^2(\gamma(\theta + \phi))$ ",
70            'data-label': "Raw data",
71            'save-name' : f'{plate_name}',
72            'loc' : 'upper right'}
73
74     tk.quick_plot_residuals(xdata, ydata, data['plotx'], data['ploty'],
75                            data['residuals'], meta=meta,
76                            uncertainty=y_unc, uncertainty_x=x_unc,
77                            save=True)
78
79
80

```

```

81     return data['chisq'], data['popt'], data['pstd']
82
83 if __name__ == '__main__':
84
85     # Part (a)
86     df = load_data('part2a.csv')
87     params = fit_plot(df, 'QWP')
88
89     printvals = [r'$\chi_{red}^2+f' = {params[0]}$']
90     for i,v in enumerate([r'I_0', r'\phi', r'\gamma']):
91         printvals.append(f'${v} = {params[1][i]} \pm {params[2][i]}$')
92
93     tk.block_print(printvals, 'Fit Parameters for QWP')
94
95     # Part (b)
96     df = load_data('part2b.csv')
97     params = fit_plot(df, 'HWP')
98
99     printvals = [r'$\chi_{red}^2+f' = {params[0]}$']
100    for i,v in enumerate([r'I_0', r'\phi', r'\gamma']):
101        printvals.append(f'${v} = {params[1][i]} \pm {params[2][i]}$')
102
103    tk.block_print(printvals, 'Fit Parameters for HWP')

```

Toolkit

```

1  # -*- coding: utf-8 -*-
2  """
3  Created on Tue Jan 23 12:34:34 2024
4  Updated on Mon Oct 14 21:22:09 2024
5  Updated on Mon Feb 10 09:51:03 2025
6
7  Lab Toolkit
8
9  @author: Aditya K. Rao
10 @github: @adiraio-projects
11 """
12 import numpy as np
13 from scipy.optimize import curve_fit
14 import matplotlib.pyplot as plt
15 import matplotlib.gridspec as gridspec
16 import os
17 import math
18 import textwrap
19
20 #from uncertainties import ufloat
21
22 font = {'family' : 'DejaVu Sans',
23        'size'    : 30}
24
25 plt.rc('font', **font)

```

```

26
27 def curve_fit_data(xdata, ydata, fit_type, override=False,
28                   override_params=(None,), uncertainty=None,
29                   res=False, chi=False, uncertainty_x=None,
30                   model_function_custom=None, guess=None):
31
32     def chi_sq_red(measured_data:list[float], expected_data:list[float],
33                   uncertainty:list[float], v: int):
34         if type(uncertainty)==float:
35             uncertainty = [uncertainty]*len(measured_data)
36         chi_sq = 0
37
38         # Converting summation in equation into a for loop
39         for i in range(0, len(measured_data)):
40             chi_sq += (pow((measured_data[i] \
41                           - expected_data[i]),2)/(uncertainty[i]**2))
42
43         chi_sq = (1/v)*chi_sq
44
45         return chi_sq
46
47
48     def residual_calculation(y_data: list, exp_y_data) -> list[float]:
49         residuals = []
50         for v, u in zip(y_data, exp_y_data):
51             residuals.append(u-v)
52
53         return residuals
54
55     def model_function_linear_int(x, m, c):
56         return m*x+c
57
58     def model_function_exp(x, a, b, c):
59         return a*np.exp**(b*x)
60
61     def model_function_log(x, a, b):
62         return b*np.log(x+a)
63
64     def model_function_linear_int_mod(x, m, c):
65         return m*(x+c)
66
67     def model_function_linear(x, m):
68         return m*x
69
70     def model_function_xlnx(x, a, b, c):
71         return b*x*(np.log(x)) + c
72
73     def model_function_lnx(x, a, b, c):
74         return b*(np.log(x)) + c
75
76     def model_function_sqrt(x, a):
77         return a*np.sqrt(x)
78
79     model_functions = {

```

```
80         'linear' : model_function_linear ,
81         'linear-int' : model_function_linear_int ,
82         'xlnx' : model_function_xlnx ,
83         'log' : model_function_log ,
84         'exp' : model_function_exp ,
85         'custom' : model_function_custom
86     }
87
88     try:
89         model_func = model_functions[fit_type]
90
91     except:
92         raise ValueError(f'Unsupported fit-type: {fit_type}')
93
94
95     if not override:
96         new_xdata = np.linspace(min(xdata), max(xdata), num=100)
97
98
99         if type(uncertainty) == int:
100             abs_sig = True
101         else:
102             abs_sig = False
103
104         if guess is not None:
105             popt, pcov = curve_fit(model_func, xdata, ydata, sigma=uncertainty,
106                                   maxfev=20000, absolute_sigma=abs_sig, p0=guess)
107         else:
108             popt, pcov = curve_fit(model_func, xdata, ydata, sigma=uncertainty,
109                                   maxfev=20000, absolute_sigma=abs_sig)
110         param_num = len(popt)
111
112         exp_ydata = model_func(xdata,*popt)
113
114         deg_free = len(xdata) - param_num
115
116         new_ydata = model_func(new_xdata, *popt)
117
118         residuals = None
119         chi_sq = None
120
121         if res:
122             residuals = residual_calculation(exp_ydata, ydata)
123
124         if chi:
125             chi_sq = chi_sq_red(ydata, exp_ydata, uncertainty, deg_free)
126
127         data_output = {
128             'popt' : popt,
129             'pcov' : pcov,
130             'plotx': new_xdata,
131             'ploty': new_ydata,
132             'chisq' : chi_sq,
133             'residuals' : residuals,
```

```

134         'pstd' : np.sqrt(np.diag(pcov))
135     }
136
137     return data_output
138
139 else:
140     return model_func(xdata, *override_params)
141
142
143 def quick_plot_residuals(xdata, ydata, plot_x, plot_y,
144                          residuals, meta=None, uncertainty=[], save=False,
145                          uncertainty_x=[]):
146     """
147     Relies on the python uncertainties package to function as normal, however,
148     this can be overridden by providing a list for the uncertainties.
149     """
150     fig = plt.figure(figsize=(14,14))
151     gs = gridspec.GridSpec(ncols=11, nrows=11, figure=fig)
152     main_fig = fig.add_subplot(gs[:6,:])
153     res_fig = fig.add_subplot(gs[8,:])
154
155     main_fig.grid('on')
156     res_fig.grid('on')
157     if type(uncertainty) is int:
158         uncertainty = [uncertainty]*len(xdata)
159
160     elif len(uncertainty) == 0:
161         for y in ydata:
162             uncertainty.append(y.std_dev)
163
164     if meta is None:
165         meta = {'title' : 'INSERT-TITLE',
166                'xlabel' : 'INSERT-XLABEL',
167                'ylabel' : 'INSERT-YLABEL',
168                'chisq' : 0,
169                'fit-label': "Best Fit",
170                'data-label': "Data",
171                'save-name' : 'IMAGE',
172                'loc' : 'lower right'}
173
174     main_fig.set_title(meta['title'], fontsize = 46)
175     if len(uncertainty_x)==0:
176         main_fig.errorbar(xdata, ydata, yerr=uncertainty, #xerr=uncertainty_x,
177                           markersize='4', fmt='o', color='red',
178                           label=meta['data-label'], ecolor='black')
179     else:
180         main_fig.errorbar(xdata, ydata, yerr=uncertainty, xerr=uncertainty_x,
181                           markersize='4', fmt='o', color='red',
182                           label=meta['data-label'], ecolor='black')
183
184     main_fig.plot(plot_x, plot_y, linestyle='dashed',
185                  label=meta['fit-label'])
186
187     main_fig.set_xlabel(meta['xlabel'])

```

```

188     main_fig.set_ylabel(meta['ylabel'])
189     main_fig.legend(loc=meta['loc'])
190
191
192     res_fig.errorbar(xdata, residuals, markersize='3', color='red', fmt='o',
193                     yerr=uncertainty, ecolor='black', alpha=0.7)
194     res_fig.axhline(y=0, linestyle='dashed', color='blue')
195     res_fig.set_title('Residuals')
196     save_name = meta["save-name"]
197     plt.savefig(f'figures/{save_name}.png')
198
199 def quick_plot_test(xdata, ydata, plot_x = [], plot_y = [],
200                     uncertainty=[]):
201     plt.figure(figsize=((14,10)))
202
203     plt.title("Test Plot for data")
204     plt.xlabel("X Data")
205     plt.ylabel("Y Data")
206
207     if len(uncertainty) != 0:
208         plt.errorbar(xdata, ydata, yerr=uncertainty, fmt='o')
209     else:
210         plt.scatter(xdata, ydata)
211
212     plt.grid("on")
213     plt.show()
214     plt.savefig('Test.png')
215     plt.close()
216
217 def block_print(data: list[str], title: str, delimiter='=') -> None:
218     """
219     Prints a formatted block of text with a title and delimiter
220
221     Parameters
222     -----
223     data : list[str]
224         Text to be printed (should be input as one block of text).
225     title : str
226         Title of the data being output.
227     delimiter : str, optional
228         Delimiter to be used. The default is '='.
229
230     Returns
231     -----
232     None.
233
234     Examples
235     -----
236     >>> r_log = 100114.24998718781
237     >>> r_dec = 0.007422298127465114
238     >>> data = [f'r^2 value (log): {r_log}',
239                 f'r^2 value (real): {r_dec}']
240     >>> block_print(data, 'Regression Coefficient', '=')
241     ===== Regression Coefficient =====

```



```

242     r^2 value (log): 100114.24998718781
243     r^2 value (real): 0.007422298127465114
244     =====
245     """
246     term_size = os.get_terminal_size().columns
247
248     breaks = 1
249     str_len = len(title)+2
250     while str_len >= term_size:
251         breaks += 1
252         str_len = math.ceil(str_len/2)
253
254
255     str_chunk_len = math.ceil(len(title)/breaks)
256     str_chunks = textwrap.wrap(title, str_chunk_len)
257     output = ''
258     for chunk in str_chunks:
259         border = delimiter*(math.floor((term_size - str_chunk_len)/2)-1)
260         output = f'{border} {chunk} {border}\n'
261
262     output=output[:-1]
263
264     output+= '\n'+ '\n'.join(data) + '\n'
265     output+=delimiter*term_size
266
267     print(output)
268
269 def numerical_methods(method_type, args=None, custom_method=None):
270     def gaussxw(N):
271
272         # Initial approximation to roots of the Legendre polynomial
273         a = np.linspace(3,4*N-1,N)/(4*N+2)
274         x = np.cos(np.pi*a+1/(8*N*N*np.tan(a)))
275
276         # Find roots using Newton's method
277         epsilon = 1e-15
278         delta = 1.0
279         while delta>epsilon:
280             p0 = np.ones(N,float)
281             p1 = np.copy(x)
282             for k in range(1,N):
283                 p0,p1 = p1,((2*k+1)*x*p1-k*p0)/(k+1)
284                 dp = (N+1)*(p0-x*p1)/(1-x*x)
285                 dx = p1/dp
286                 x -= dx
287                 delta = max(abs(dx))
288
289         # Calculate the weights
290         w = 2*(N+1)*(N+1)/(N*N*(1-x*x)*dp*dp)
291
292         return x, w
293
294     def gaussxwab(N,a,b):
295         x,w = gaussxw(N)

```

```
296         return 0.5*(b-a)*x+0.5*(b+a),0.5*(b-a)*w
297
298     methods = {
299         'gausswx' : gaussxw,
300         'gaussxwab' : gaussxwab,
301         'custom' : custom_method
302     }
303
304     try:
305         method = methods[method_type]
306
307     except:
308         raise ValueError(f'Unsupported method-type: {method_type}')
309
310     return method(*args)
311
312
313 def interpolation_methods(method_type, args=None, custom_method=None):
314
315     methods = {
316         'custom' : custom_method
317     }
318
319     try:
320         method = methods[method_type]
321
322     except:
323         raise ValueError(f'Unsupported method-type: {method_type}')
324
325     return method(*args)
```
