```
1 import pandas as pd
       2 import numpy as np
       3 import matplotlib.pyplot as plt
       1 # Read the CSV file
       2 #df = pd.read_csv('/content/adv lab 1 - Sheet1.csv')
       3 df = pd.read_csv('/content/adv lab 1 - 10_22 data.csv')
       5 # Convert the first three columns to NumPy arrays
       6 # Adjust column indices if you need different columns
       7 current = df.iloc[:, 0].values
       8 f_lo = df.iloc[:, 1].values
       9 f_hi = df.iloc[:, 2].values
      11 print(current)
      12 print(f_lo)
      13 print(f_hi)
     [-2. -1.8 -1.6 -1.4 -1.2 -1. -0.8 -0.6 -0.4 -0.2 0. 0.2 0.4 0.6 0.8 1. 1.2 1.4 1.6 1.8 2.]
     [152.5879 153.3203 153.8086 154.2969 154.7852 155.0293 155.5176 155.5176
      155.5176 155.5176 155.2734 155.0293 154.541 154.0527 153.3203 152.832
      152.0996 151.3672 150.6348 149.9023 149.1699]
     [170.4102 169.6777 169.1895 168.457 167.9688 167.4805 167.4805 166.9922
      166.9922 166.9922 167.2363 167.4805 167.9688 168.457 168.9453 169.4336
      170.166 170.6543 171.3867 172.1191 172.6074]
       1 # Calculate uncertainty due to resolution
       2 resolution_uncertainty = 97.5 / 400
       4 # Calculate uncertainty due to accuracy (25 ppm of the frequency value)
       5 \ accuracy_f_lo_unc = f_lo * (25e-6)
       6 accuracy_f_hi_unc = f_hi * (25e-6)
       8 # Combine uncertainties
       9 f_lo_unc = np.sqrt(resolution_uncertainty**2 + accuracy_f_lo_unc**2)
      10 f_hi_unc = np.sqrt(resolution_uncertainty**2 + accuracy_f_hi_unc**2)
      13 print("Uncertainty for f_lo:", f_lo_unc)
      14 print("Uncertainty for f hi:", f hi unc)
     Uncertainty for f lo: [0.24377985 0.24378014 0.24378033 0.24378052 0.24378071 0.24378081
      0.24378101 \ 0.24378101 \ 0.24378101 \ 0.24378101 \ 0.24378091 \ 0.24378081
      0.24378062 0.24378042 0.24378014 0.24377994 0.24377966 0.24377937
      0.24377909 0.24377881 0.24377853]
     Uncertainty for f_hi: [0.24378723 0.24378691 0.2437867 _0.24378638 0.24378617 _0.24378596
      0.24378596 0.24378575 0.24378575 0.24378575 0.24378585 0.24378596
      0.24378617 \ \ 0.24378638 \ \ 0.24378659 \ \ 0.2437868 \ \ \ \ 0.24378712 \ \ 0.24378733
      0.24378766 0.24378798 0.24378819]
\omega_- = \sqrt{rac{\kappa}{I}} \implies \kappa = I \omega_-^2
\overline{\omega_+} = \sqrt{rac{\kappa + 2\lambda}{I}} \implies \lambda = rac{I\omega_+^2 - \kappa}{2}
        1 I = 1.29e-8
```

 $3 \text{ kappa} = I*(2*np.pi*f_lo[10])**2$

9 print("kappa:", kappa)

14 # kappa = 0.01220280

13

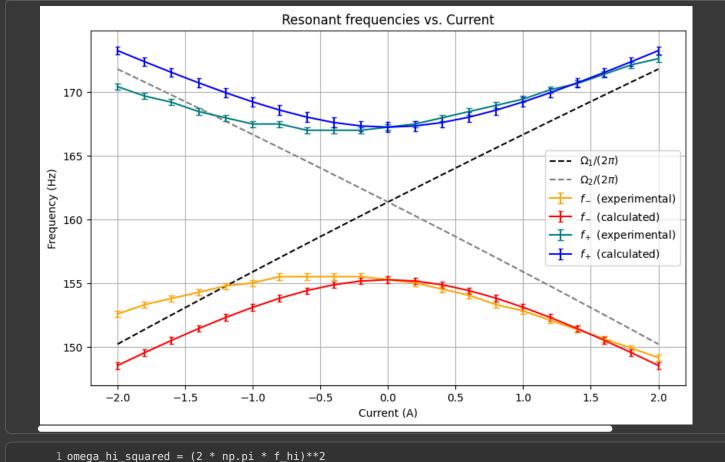
10 print("kappa_unc:", kappa_unc)
11 print("lambda_val:", lambda_val)
12 print("lambda_unc:", lambda_unc)

4 kappa_unc = 8 * np.pi**2 * I * f_lo[10] * f_lo_unc[10]

7 lambda_unc = np.sqrt((kappa_unc**2)/4 + (I*(2*np.pi*f_hi[10])*(2*np.pi*f_hi_unc[10]))**2)

6 lambda val = (I*(2*np.pi*f hi[10])**2 - kappa) / 2

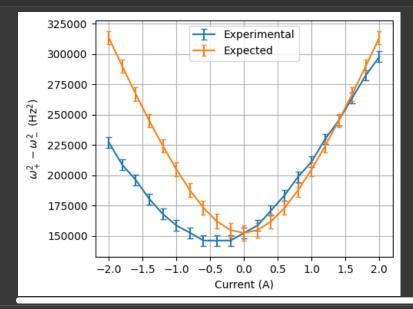
```
15 \# kappa\_unc = 2.5e-8
 17 \#lambda_val = 0.001021
 18 \# lambda unc = 0.000019
 20 \text{ mu} = 0.26
 21 B T = 3.4e-3*current
 23 f_hi_calc = np.sqrt((kappa + lambda_val + np.sqrt(lambda_val**2 + (mu * B_T)**2)) / I) / (2*np.pi)
 24 f_hi_calc_unc = np.sqrt(((lambda_unc * lambda_val / np.sqrt(B_T**2 * mu**2 + lambda_val**2) + lambda_unc)**2 +
 26 f_lo_calc = np.sqrt((kappa + lambda_val - np.sqrt(lambda_val**2 + (mu * B_T)**2)) / I) / (2*np.pi)
 27 f lo calc unc = np.sqrt((kappa unc**2 * (lambda val**2 + B T**2 * mu**2) + lambda unc**2 * (2 * lambda val**2 +
 29 print("f hi calc:", f hi calc)
 30 print("f_lo_calc:", f_lo_calc)
 31 print("f_hi_calc_unc:", f_hi_calc_unc)
 32 print("f lo calc unc:", f lo calc unc)
kappa: 0.012278450750887734
kappa unc: 3.855459948494936e-05
lambda val: 0.0009824234153908993
lambda unc: 2.8332194887643867e-05
f_hi_calc: [173.2353678 172.36851929 171.52385231 170.70919486 169.93510249
169.21575695 168.56989739 168.02133834 167.59811689 167.32892563
              167.32892563 167.59811689 168.02133834 168.56989739
 169.21575695 169.93510249 170.70919486 171.52385231 172.36851929
 173.2353678 ]
f lo calc: [148.55071904 149.55568309 150.52367546 151.44695301 152.31503448
153.11380207 153.82457046 154.42356896 154.88279439 155.1735784
             155.1735784 154.88279439 154.42356896 153.82457046
153.11380207 152.31503448 151.44695301 150.52367546 149.55568309
148.55071904]
f hi_calc_unc: [0.32350564 0.32987676 0.33699752 0.34496179 0.35382637 0.36354935
0.37388154 \ 0.38421033 \ 0.3934215 \ 0.39996686 \ 0.40235984 \ 0.39996686
0.3934215 \quad 0.38421033 \quad 0.37388154 \quad 0.36354935 \quad 0.35382637 \quad 0.34496179
0.33699752 0.32987676 0.32350564]
f_lo_calc_unc: [0.27240318 0.26805575 0.26370297 0.25941676 0.25531616 0.25157735
0.24462601 0.24607457 0.2484244 0.25157735 0.25531616 0.25941676
0.26370297 0.26805575 0.27240318]
  1 Omega1 = np.sqrt((kappa + mu*B_T + lambda_val)/I)
  2 Omega2 = np.sqrt((kappa - mu*B T + lambda val)/I)
  1 plt.rcParams['figure.figsize'] = [10, 6]
  3 plt.plot(current, Omega1/(2*np.pi),"--", label=r'$\Omega_1/(2\pi)$', color="0")
  4 plt.plot(current, Omega2/(2*np.pi),"--", label=r'$\Omega_2/(2\pi)$', color="0.5")
  6 plt.errorbar(current, f_lo, yerr=f_lo_unc, capsize=3, label='$f_-$ (experimental)', color="orange")
  7 plt.errorbar(current, f_lo_calc, yerr=f_lo_calc_unc, capsize=2, label='$f_-$ (calculated)', color="red")
  8 #plt.plot(current, f_lo_calc, label='$f_-$ (calculated)', color="red")
 11 plt.errorbar(current, f_hi, yerr=f_hi_unc, capsize=3, label='$f_+$ (experimental)', color="teal")
 12 plt.errorbar(current, f_hi_calc, yerr=f_hi_calc_unc, capsize=2, label='$f_+$ (calculated)', color="blue")
 16 plt.xlabel('Current (A)')
 17 plt.ylabel('Frequency (Hz)')
 18 plt.title('Resonant frequencies vs. Current')
 19 plt.legend()
 20 plt.grid(True)
 21 plt.savefig("avoided_crossing.svg", format="svg")
```



```
2 omega hi squared unc = 8 * np.pi**2 * f hi * f hi unc
4 \text{ omega\_lo\_squared} = (2 * np.pi * f_lo)**2
5 omega_lo_squared_unc = 8 * np.pi**2 * f_lo * f_lo_unc
7 omega_squared_diff = omega_hi_squared - omega_lo_squared
8 omega_squared_diff_unc = np.sqrt(omega_hi_squared_unc**2 + omega_lo_squared_unc**2)
10 omega_squared_sum = omega_hi_squared + omega_lo_squared
11 omega_squared_sum_unc = np.sqrt(omega_hi_squared_unc**2 + omega_lo_squared_unc**2)
1 omega_hi_squared_calc = (2 * np.pi * f_hi_calc)**2
2 omega_hi_squared_calc_unc = 8 * np.pi**2 * f_hi_calc * f_hi_calc_unc
4 omega_lo_squared_calc = (2 * np.pi * f_lo_calc)**2
5 omega_lo_squared_calc_unc = 8 * np.pi**2 * f_lo_calc * f_lo_calc_unc
7 omega squared diff calc = omega hi squared calc - omega lo squared calc
8 omega_squared_diff_calc_unc = np.sqrt(omega_hi_squared_calc_unc**2 + omega_lo_squared_calc_unc**2)
10 omega_squared_sum_calc = omega_hi_squared_calc + omega_lo_squared_calc
11 omega_squared_sum_calc_unc = np.sqrt(omega_hi_squared_calc_unc**2 + omega_lo_squared_calc_unc**2)
1 plt.rcParams['figure.figsize'] = [5, 4]
```

```
4 # Plot the experimental values with error bars
5 plt.errorbar(current, omega_squared_diff, yerr=omega_squared_diff_unc, capsize=3, label='Experimental')
6
7 # Plot the expected values
8 plt.errorbar(current, omega_squared_diff_calc, yerr=omega_squared_diff_calc_unc, capsize=2, label='Expected')
9
10 # Label the axes and add a title
11 plt.xlabel('Current (A)')
12 plt.ylabel(r'$\omega_+^2 - \omega_-^2$ (Hz$^2$)')
13 #plt.title("Difference of squared frequencies")
14
15 # Add a legend and grid
16 plt.legend()
```

```
17 plt.grid(True)
18
19 # Display the plot
20 plt.savefig("omegadiff.svg", format="svg")
```



```
1 # Plot the experimental values with error bars
2 plt.errorbar(current, omega_squared_sum, yerr=omega_squared_sum_unc, capsize=3, label='Experimental')
3
4 # Plot the expected values
5 plt.errorbar(current, omega_squared_sum_calc, yerr=omega_squared_sum_calc_unc, capsize=2, label='Expected')
6
7 # Label the axes and add a title
8 plt.xlabel('Current (A)')
9 plt.ylabel(r'$\omega_+^2 + \omega_-^2$ (Hz$^2$)')
10 #plt.title("Sum of squared frequencies")
11
12 # Add a legend and grid
13 plt.legend()
14 plt.grid(True)
15
16 # Display the plot
17 plt.savefig("omegasum.svg", format="svg")
```

