

```

# imports
import numpy as np
import matplotlib.pyplot as plt
from scipy.interpolate import interp1d
import scipy.optimize as opt
import os

user_path = '/Users/oliviapereira/Desktop' if
(os.path.isdir('/Users/oliviapereira')) else
'/Users/audrey/Documents/PHYS469'

```

Load in data

```

folder_path = user_path +
'/PHYS469/saturation_spectroscopy/raw_data/10-31-2024/'
raw_probe_pump_data = {}
raw_probe_data = {}

for file_name in os.listdir(folder_path):
    file_name = file_name.strip()
    # Read the lines of the file
    with open(folder_path+file_name, 'r') as f:
        lines = f.readlines()
    # Find end of the header (the line that starts with 'x ys')
    for i, line in enumerate(lines):
        if line.startswith('x    y'):
            lines_to_skip = i
            break

    # Read data (skip header)
    data = np.genfromtxt(folder_path+file_name,
skip_header=lines_to_skip, names=True, missing_values='_',
filling_values=0)
    reshaped_data = np.vstack([data[field] for field in
data.dtype.names]).T

    # Make a dict for the run
    file_dict = {label: reshaped_data[:,i] for i,label in
enumerate(data.dtype.names)}
    # Add dict to data dict
    if 'run' in file_name:
        raw_probe_pump_data[file_name[:-4]] = file_dict
    else:
        raw_probe_data[file_name[:-4]] = file_dict

# showing keys for example
raw_probe_pump_data.keys(), raw_probe_data.keys(),
raw_probe_pump_data['run1'].keys()

```

```
(dict_keys(['run20', 'run9', 'run8', 'run19', 'run18', 'run6',
'run15', 'run14', 'run7', 'run5', 'run16', 'run17', 'run4', 'run13',
'run12', 'run1', 'run3', 'run10', 'run11', 'run2'])),
dict_keys(['nopump17', 'nopump16', 'nopump14', 'nopump15',
'nopump11', 'nopump10', 'nopump12', 'nopump13', 'nopump2', 'nopump3',
'nopump1', 'nopump4', 'nopump5', 'nopump7', 'nopump6', 'nopump8',
'nopump9', 'nopump20', 'nopump18', 'nopump19'])),
dict_keys(['x', 'y2', 'y3']))
```

Identifying peaks in hyperfine runs

```
i = 1
# loading run
probe_data = raw_probe_pump_data['run'+str(i)]['y2'] -
raw_probe_data['nopump1']['y2']

plt.figure(figsize = (10, 3), dpi = 200)
plt.plot(np.arange(len(probe_data)), probe_data, zorder = 0, marker =
'.', markersize = 1, alpha = 0.5)

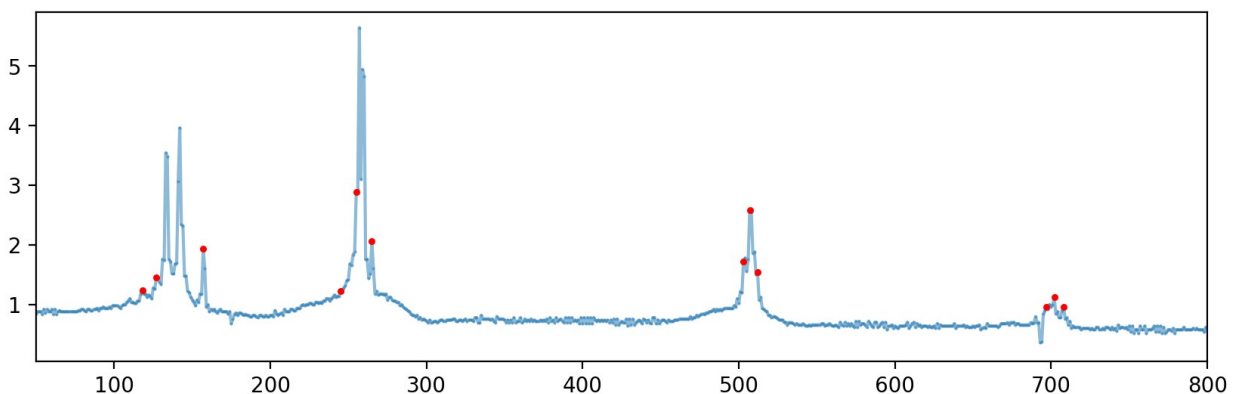
# choosing indices of peaks
indices = [118, 127, 157,      245, 255, 265,      503, 507, 512,      697,
702, 708]

# plotting these points from the original data
plt.scatter(indices, probe_data[indices], color='r', s = 5)

plt.xlim(50, 800)

# to zoom in on specific peak
# plt.xlim(115, 165)
# plt.xlim(240, 270)
# plt.xlim(495, 520)
# plt.xlim(685, 720)

(50.0, 800.0)
```



```
# FINDING THE HYPERFINE PEAK INDICES BY HAND FOR EACH RUN :(
hyperfine_peaks = {
    'run1': [118, 127, 157, 245, 255, 265, 503, 507, 512,
697, 702, 708],
    'run2': [118, 127, 157, 245, 255, 265, 503, 507, 512,
697, 702, 708],
    'run3': [118, 127, 157, 245, 255, 264, 503, 507, 512,
697, 702, 708],
    'run4': [118, 127, 157, 245, 253, 264, 503, 507, 512,
697, 702, 708],
    'run5': [118, 127, 156, 245, 253, 264, 503, 507, 512,
697, 702, 708],
    'run6': [118, 127, 156, 245, 253, 264, 503, 507, 512,
697, 702, 708],
    'run7': [118, 126, 156, 245, 253, 264, 503, 507, 512,
696, 702, 708],
    'run8': [118, 126, 156, 245, 253, 264, 503, 507, 512,
696, 702, 708],
    'run9': [117, 126, 156, 245, 253, 264, 503, 507, 512,
696, 702, 708],
    'run10': [118, 126, 156, 245, 253, 264, 503, 507, 512,
696, 702, 708],
    'run11': [111, 121, 150, 245, 253, 263, 503, 507, 512,
698, 703, 709],
    'run12': [109, 119, 149, 245, 253, 263, 503, 507, 512,
698, 703, 709],
    'run13': [109, 118, 148, 245, 253, 263, 503, 507, 512,
698, 703, 709],
    'run14': [109, 118, 148, 245, 253, 263, 503, 507, 512,
698, 704, 710],
    'run15': [109, 117, 147, 243, 251, 263, 503, 507, 512,
698, 704, 710],
    'run16': [108, 117, 147, 243, 251, 263, 503, 507, 512,
698, 704, 710],
    'run17': [108, 117, 147, 243, 251, 263, 503, 507, 512,
698, 704, 710],
    'run18': [107, 115, 146, 243, 251, 262, 501, 505, 512,
698, 703, 709],
    'run19': [107, 115, 146, 243, 251, 262, 501, 503, 512,
698, 703, 709],
    'run20': [107, 115, 146, 243, 251, 262, 501, 503, 512,
698, 703, 709],
}
```

Defining fit for interferometer data

$$I(t) = A + A \cos\left(\frac{4\pi \Delta L}{c} f(t)\right) + g(t)$$

where $f(t)$ is the frequency sweep, and $g(t)$ is the ramp, with

$$f(t) = f_0 + \alpha t + \beta t^2$$

and

$$g(t) = a + b t$$

```
def interference_model(t, f0, alpha, A, a, b, beta):
    delta_L = 0.34      # path difference in m
    c = 3e8             # speed of light in m/s
    f_t = f0 + alpha * t + beta*t**2      # quadratic model for
the frequency sweep f(t)
    phase = (4 * np.pi * delta_L / c) * f_t      # phase expression
from lab manual
    ramp = a + b*t      # linear model for the
ramp g(t)
    return A * (1 + np.cos(phase)) + ramp
```

Converting all runs to frequency space

```
hyperfine_freq_gaps = {}
ground_freq_gaps = {} # 87Rb (F=1, F'=2); 85Rb (F=2, F'=3); 85Rb
(F=3, F'=3); 87Rb (F=2, F'=2)
ground_gaps_indices = [1,4,8,11]
parameters = []
for run in hyperfine_peaks.keys():

    # load in hyperfine peak locations
    hyperfine_indices = hyperfine_peaks[run]

    # load in data and trim
    inter_data = raw_probe_pump_data[run]['y3'][:800]
    time = raw_probe_pump_data[run]['x'][:800] / 10 # divide by 10 to
convert to seconds

    # fit the interferometer data to the function
    alpha_guess = 4e11 if (run[-1]==9) else 4.4e11
    initial_guess = [1e14, alpha_guess, 1, 4.5, -100, 1e12] # f0,
alpha, A, a, b, beta

    params, covariance = opt.curve_fit(interference_model, time[100:],
inter_data[100:], p0=initial_guess, maxfev=1000000000)
    errors = np.sqrt(np.diag(covariance))

    if run == 'run1': parameters = params

    print(f'f0={params[0]:.2e} ± {errors[0]:.2e}\n',
          f'alpha={params[1]:.2e} ± {errors[1]:.2e}\n',
          f'A={params[2]:.2e} ± {errors[2]:.2e}\n',
```

```

f'a={params[3]:.2e} ± {errors[3]:.2e}\n',
f'b={params[4]:.2e} ± {errors[4]:.2e}\n',
f'beta={params[5]:.2e} ± {errors[5]:.2e}')

# plot the fit
fig, axes = plt.subplots(2, 1, figsize=(5,4), dpi=150,
gridspec_kw={'height_ratios': [2, 1]})
fig.suptitle(f'{run}')
axes[0].plot(time, interference_model(time, *params), color = 'r')
axes[0].plot(time, inter_data, color = 'k')
axes[1].scatter(time, inter_data-interference_model(time,
*params), color = 'r', s=0.5)
plt.show()

# create the frequency function for this run
def f(t, f0, alpha, beta):
    return f0 + alpha * t + beta * t**2

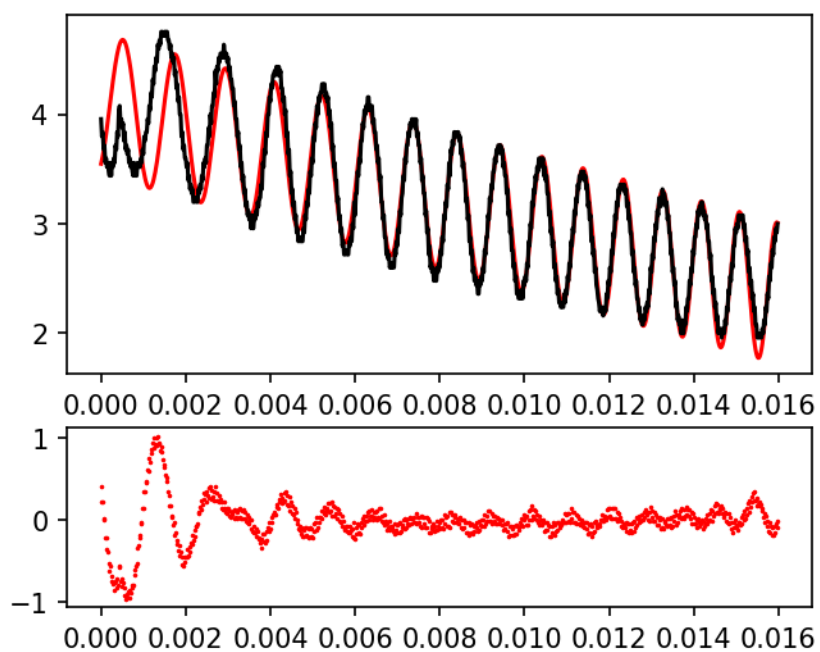
# convert hyperfine peaks to frequency
hyperfine_peaks_time = hyperfine_peaks[run]
hyperfine_peak_freqs = f(time[hyperfine_peaks_time], params[0],
params[1], params[5])
ground_gaps_freqs = hyperfine_peak_freqs[ground_gaps_indices] #
freqs to calculate the ground state gaps

# calculate the frequency gaps
hyperfine_freq_gaps[run] = np.diff(hyperfine_peak_freqs)
ground_freq_gaps[run] = [np.abs(ground_gaps_freqs[1]-
ground_gaps_freqs[2]),
np.abs(ground_gaps_freqs[0]-
ground_gaps_freqs[-1]))] # 85Rb, 87Rb

f0=1.00e+14 ± 3.65e+06
alpha=3.47e+11 ± 9.15e+08
A=6.45e-01 ± 6.75e-03
a=3.45e+00 ± 1.32e-02
b=-1.08e+02 ± 1.18e+00
beta=4.92e+12 ± 5.01e+10

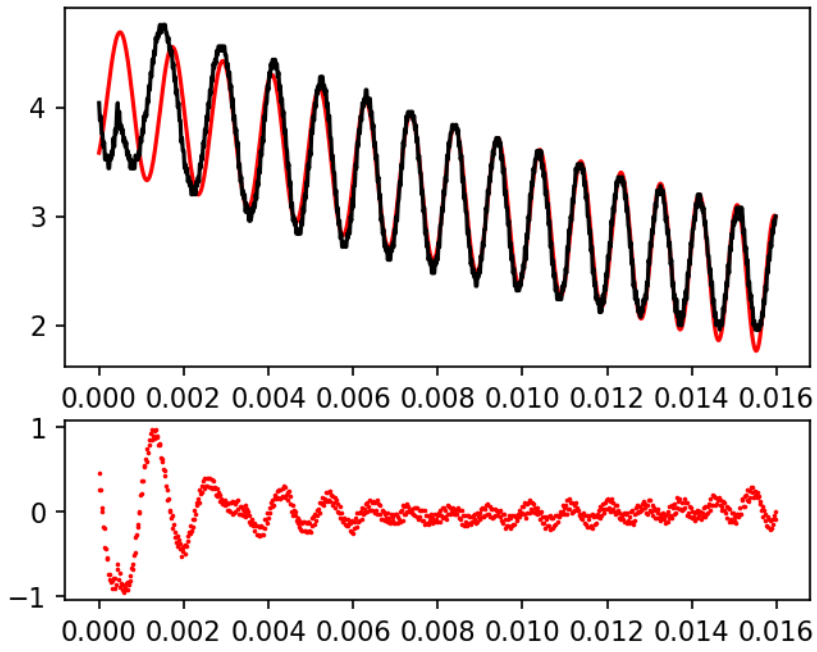
```

run1



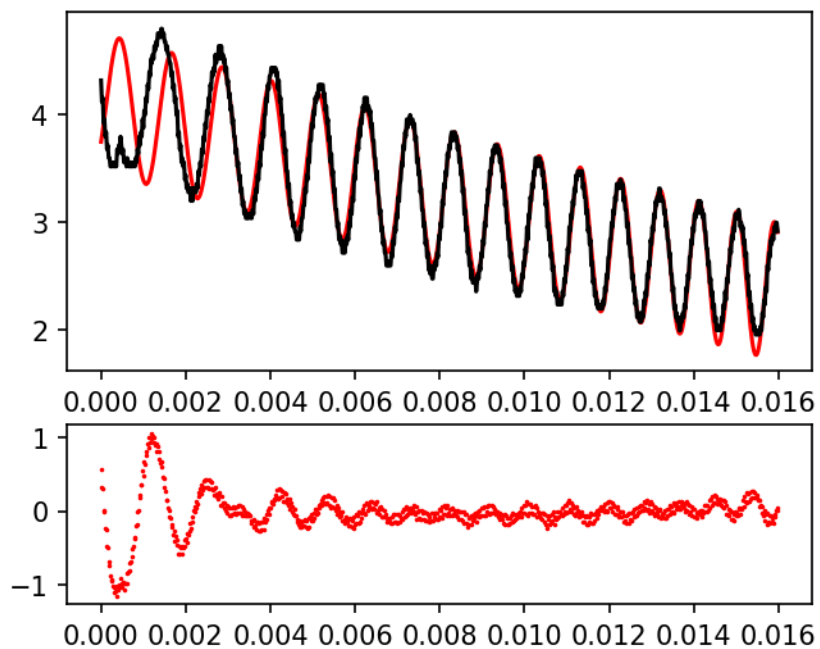
$f_0=1.00\text{e}+14 \pm 3.66\text{e}+06$
 $\alpha=3.47\text{e}+11 \pm 9.16\text{e}+08$
 $A=6.45\text{e}-01 \pm 6.75\text{e}-03$
 $a=3.45\text{e}+00 \pm 1.32\text{e}-02$
 $b=-1.09\text{e}+02 \pm 1.18\text{e}+00$
 $\beta=4.95\text{e}+12 \pm 5.02\text{e}+10$

run2



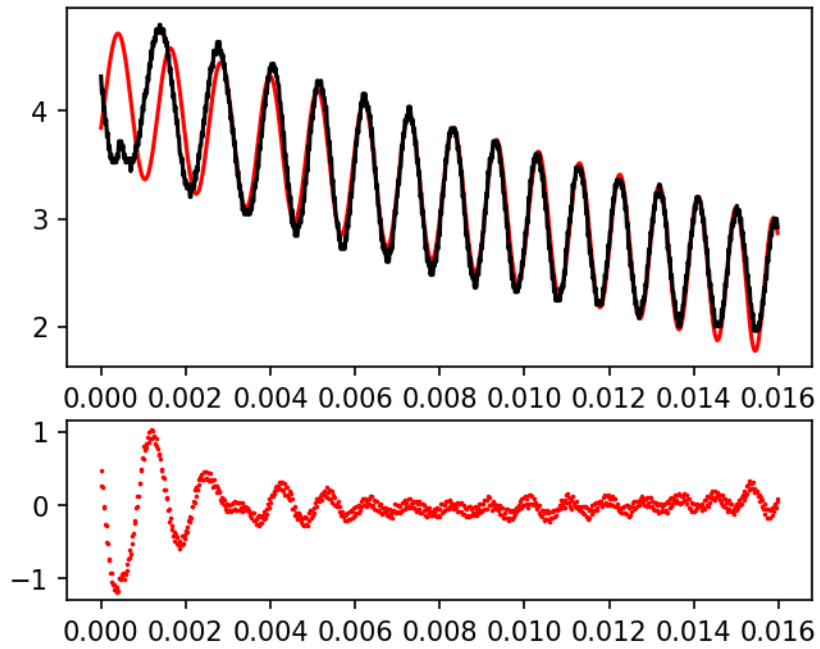
$f_0=1.00\text{e}+14 \pm 3.64\text{e}+06$
 $\alpha=3.48\text{e}+11 \pm 9.09\text{e}+08$
 $A=6.43\text{e}-01 \pm 6.54\text{e}-03$
 $a=3.47\text{e}+00 \pm 1.28\text{e}-02$
 $b=-1.10\text{e}+02 \pm 1.15\text{e}+00$
 $\beta=4.91\text{e}+12 \pm 4.97\text{e}+10$

run3



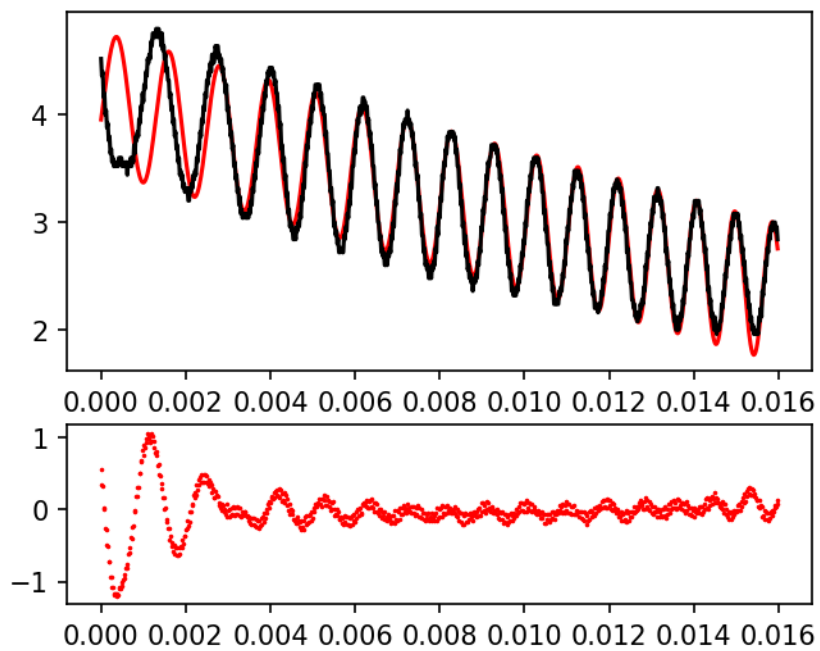
$f_0=1.00\text{e}+14 \pm 3.82\text{e}+06$
 $\alpha=3.47\text{e}+11 \pm 9.52\text{e}+08$
 $A=6.43\text{e}-01 \pm 6.80\text{e}-03$
 $a=3.48\text{e}+00 \pm 1.33\text{e}-02$
 $b=-1.10\text{e}+02 \pm 1.20\text{e}+00$
 $\beta=4.93\text{e}+12 \pm 5.20\text{e}+10$

run4



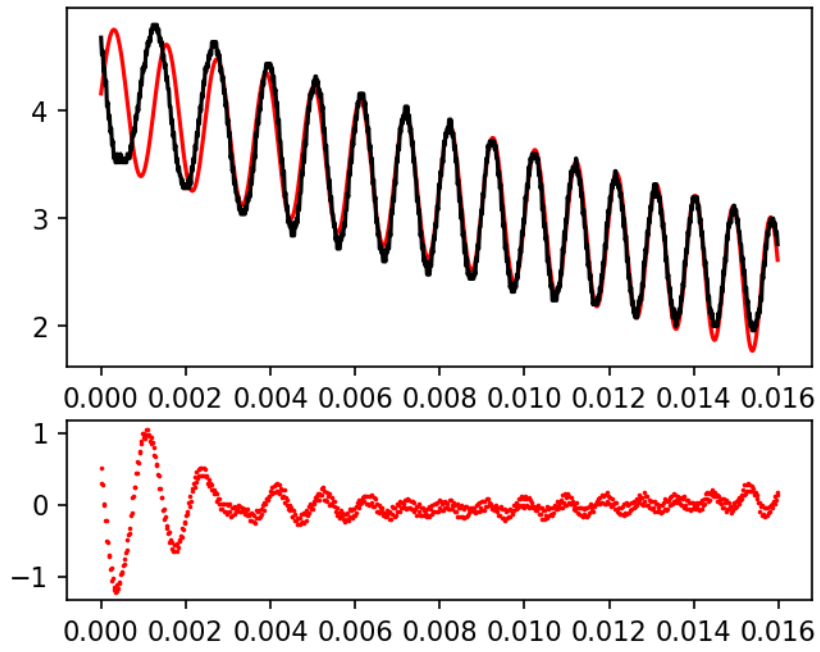
$f_0=1.00\text{e}+14 \pm 3.80\text{e}+06$
 $\alpha=3.48\text{e}+11 \pm 9.44\text{e}+08$
 $A=6.43\text{e}-01 \pm 6.70\text{e}-03$
 $a=3.48\text{e}+00 \pm 1.31\text{e}-02$
 $b=-1.11\text{e}+02 \pm 1.18\text{e}+00$
 $\beta=4.88\text{e}+12 \pm 5.15\text{e}+10$

run5



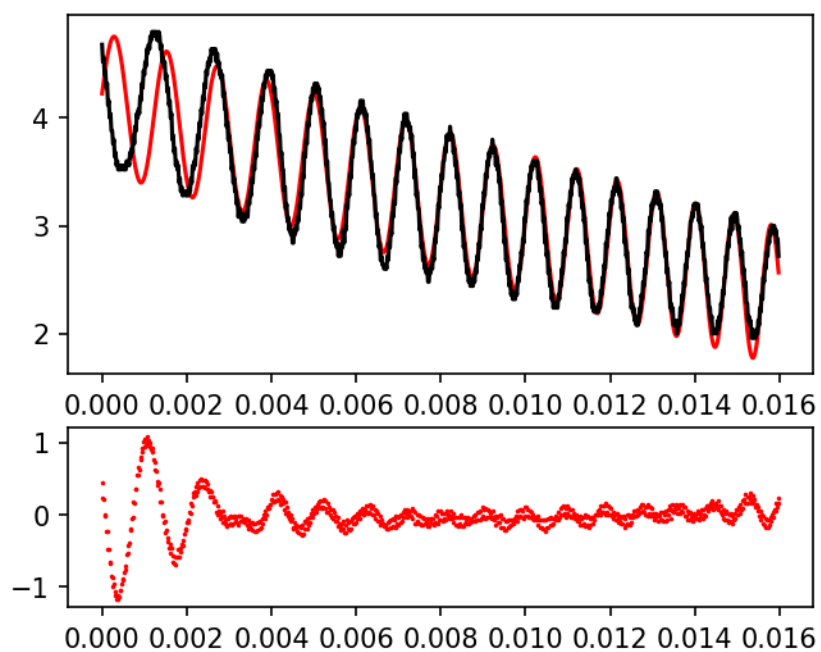
$f_0=1.00\text{e}+14 \pm 3.88\text{e}+06$
 $\alpha=3.48\text{e}+11 \pm 9.60\text{e}+08$
 $A=6.46\text{e}-01 \pm 6.82\text{e}-03$
 $a=3.50\text{e}+00 \pm 1.34\text{e}-02$
 $b=-1.12\text{e}+02 \pm 1.20\text{e}+00$
 $\beta=4.90\text{e}+12 \pm 5.22\text{e}+10$

run6



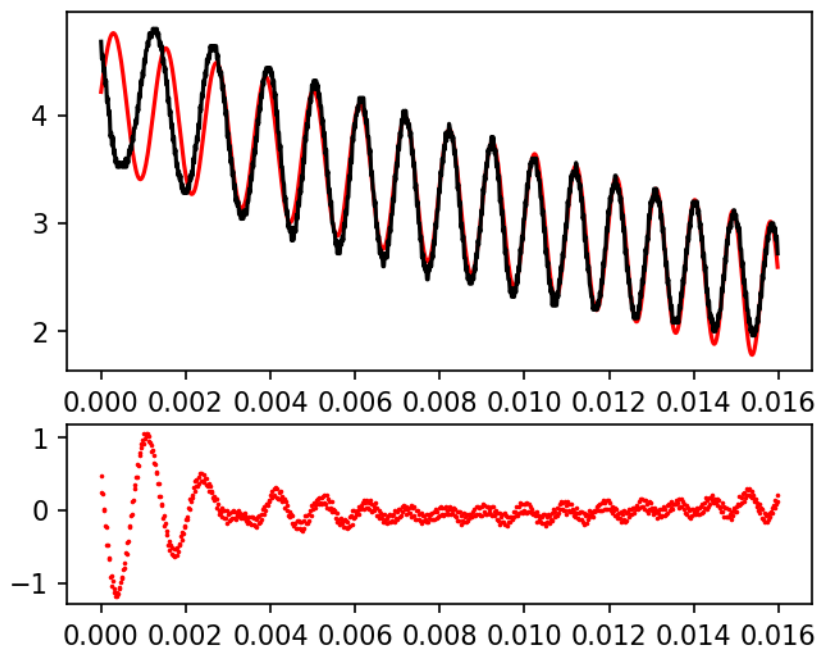
$f_0=1.00\text{e}+14 \pm 3.88\text{e}+06$
 $\alpha=3.48\text{e}+11 \pm 9.60\text{e}+08$
 $A=6.43\text{e}-01 \pm 6.79\text{e}-03$
 $a=3.50\text{e}+00 \pm 1.34\text{e}-02$
 $b=-1.12\text{e}+02 \pm 1.19\text{e}+00$
 $\beta=4.90\text{e}+12 \pm 5.22\text{e}+10$

run7



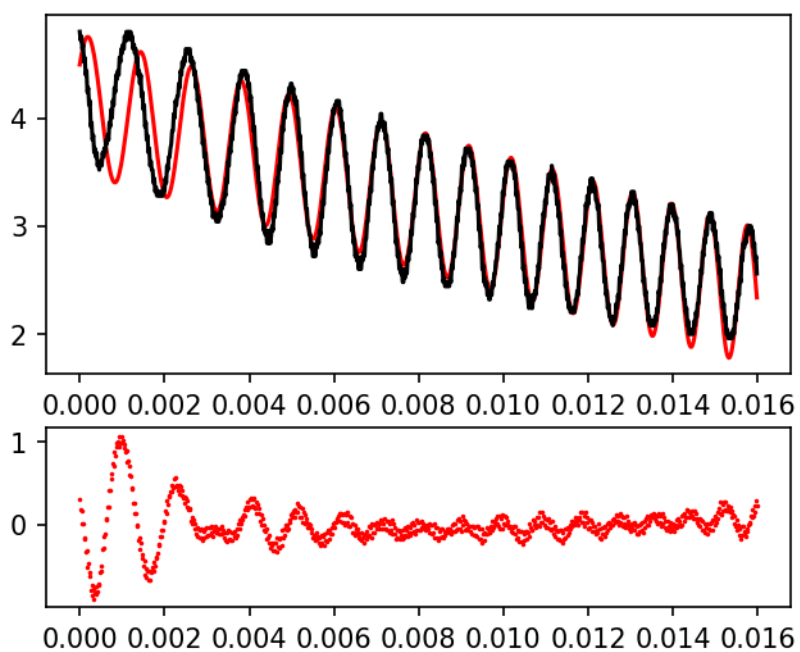
$f_0=1.00\text{e}+14 \pm 3.88\text{e}+06$
 $\alpha=3.47\text{e}+11 \pm 9.60\text{e}+08$
 $A=6.42\text{e}-01 \pm 6.77\text{e}-03$
 $a=3.51\text{e}+00 \pm 1.33\text{e}-02$
 $b=-1.12\text{e}+02 \pm 1.19\text{e}+00$
 $\beta=4.91\text{e}+12 \pm 5.22\text{e}+10$

run8



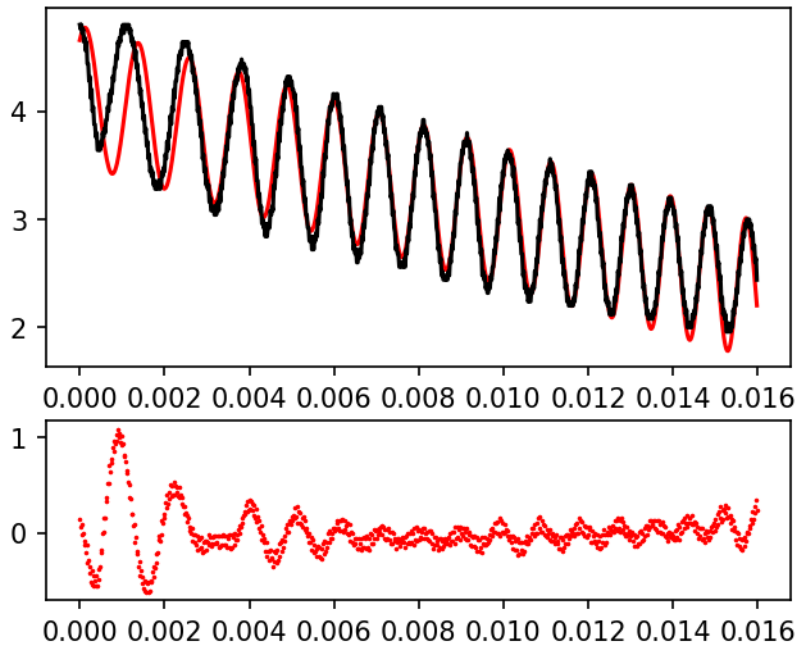
$f_0=1.00\text{e}+14 \pm 4.06\text{e}+06$
 $\alpha=3.46\text{e}+11 \pm 1.00\text{e}+09$
 $A=6.40\text{e}-01 \pm 7.18\text{e}-03$
 $a=3.49\text{e}+00 \pm 1.42\text{e}-02$
 $b=-1.12\text{e}+02 \pm 1.26\text{e}+00$
 $\beta=4.94\text{e}+12 \pm 5.45\text{e}+10$

run9



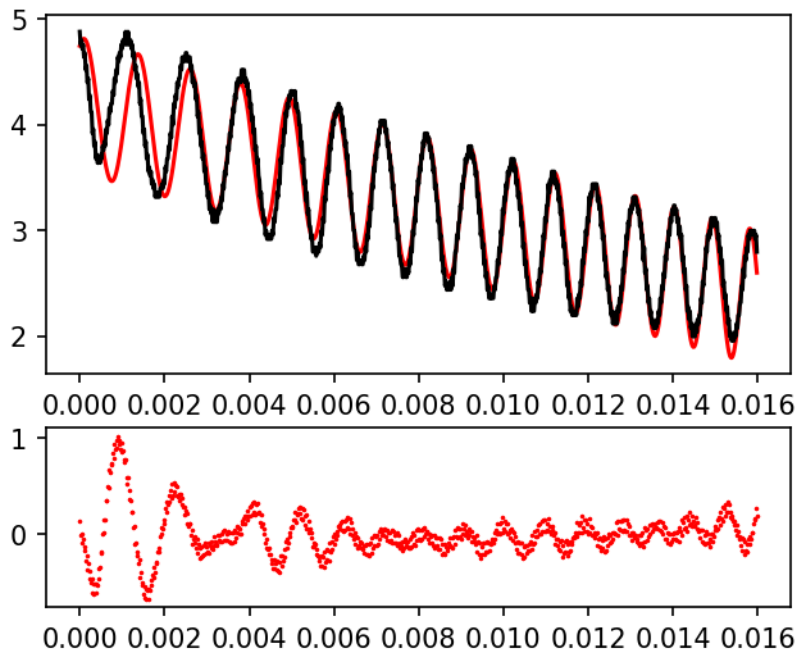
$f_0=1.00\text{e}+14 \pm 4.02\text{e}+06$
 $\alpha=3.45\text{e}+11 \pm 9.96\text{e}+08$
 $A=6.40\text{e}-01 \pm 7.25\text{e}-03$
 $a=3.51\text{e}+00 \pm 1.44\text{e}-02$
 $b=-1.13\text{e}+02 \pm 1.27\text{e}+00$
 $\beta=4.97\text{e}+12 \pm 5.41\text{e}+10$

run10



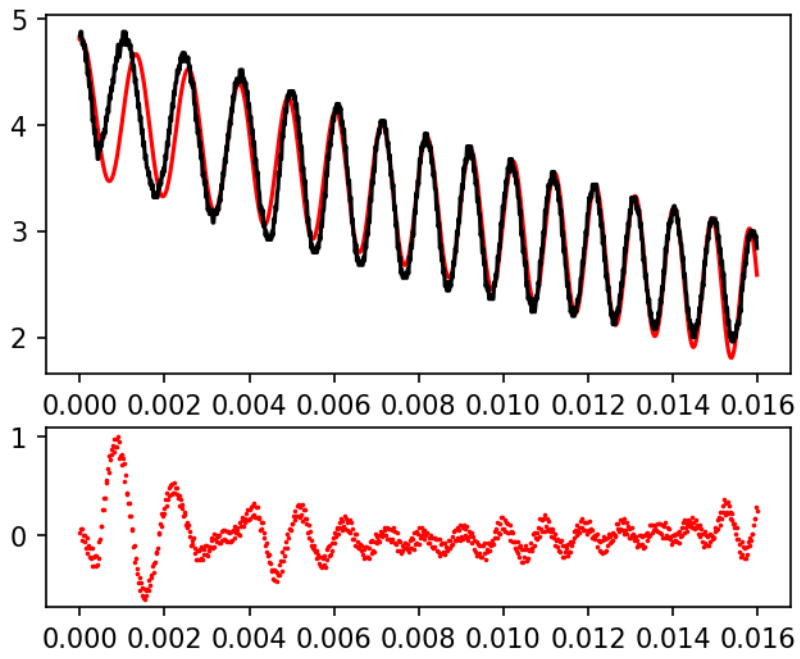
$f_0=1.00\text{e}+14 \pm 4.43\text{e}+06$
 $\alpha=3.38\text{e}+11 \pm 1.10\text{e}+09$
 $A=6.36\text{e}-01 \pm 7.86\text{e}-03$
 $a=3.55\text{e}+00 \pm 1.57\text{e}-02$
 $b=-1.14\text{e}+02 \pm 1.38\text{e}+00$
 $\beta=5.25\text{e}+12 \pm 6.01\text{e}+10$

run11



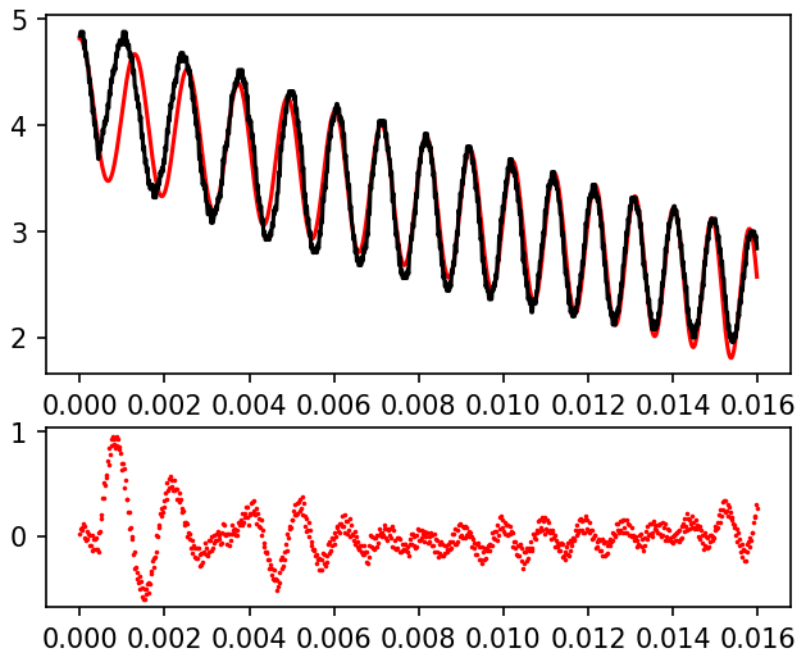
$f_0=1.00\text{e}+14 \pm 4.58\text{e}+06$
 $\alpha=3.35\text{e}+11 \pm 1.14\text{e}+09$
 $A=6.35\text{e}-01 \pm 8.24\text{e}-03$
 $a=3.55\text{e}+00 \pm 1.64\text{e}-02$
 $b=-1.13\text{e}+02 \pm 1.44\text{e}+00$
 $\beta=5.36\text{e}+12 \pm 6.25\text{e}+10$

run12



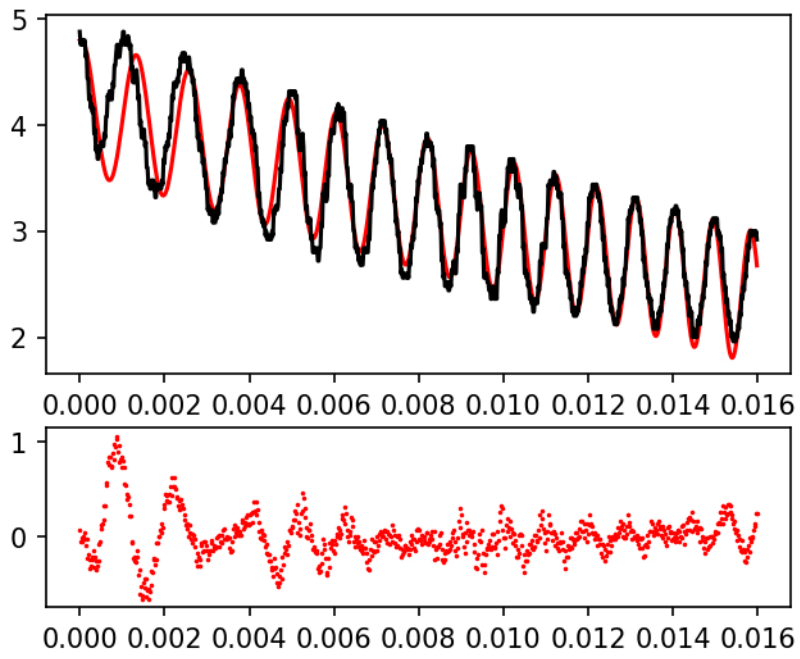
$f_0=1.00\text{e}+14 \pm 4.67\text{e}+06$
 $\alpha=3.33\text{e}+11 \pm 1.17\text{e}+09$
 $A=6.34\text{e}-01 \pm 8.45\text{e}-03$
 $a=3.55\text{e}+00 \pm 1.69\text{e}-02$
 $b=-1.13\text{e}+02 \pm 1.48\text{e}+00$
 $\beta=5.40\text{e}+12 \pm 6.39\text{e}+10$

run13



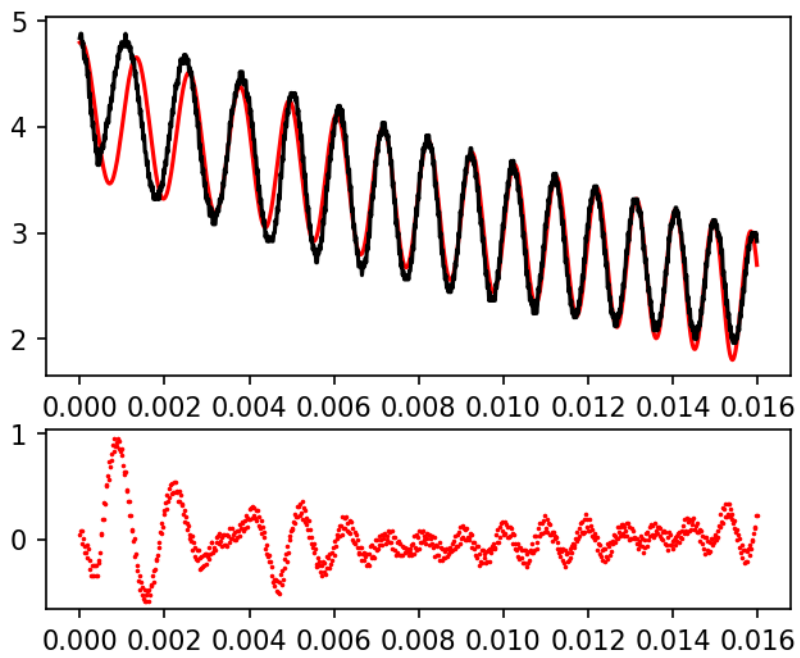
$f_0=1.00\text{e}+14 \pm 5.07\text{e}+06$
 $\alpha=3.34\text{e}+11 \pm 1.27\text{e}+09$
 $A=6.27\text{e}-01 \pm 8.97\text{e}-03$
 $a=3.56\text{e}+00 \pm 1.79\text{e}-02$
 $b=-1.14\text{e}+02 \pm 1.57\text{e}+00$
 $\beta=5.38\text{e}+12 \pm 6.93\text{e}+10$

run14



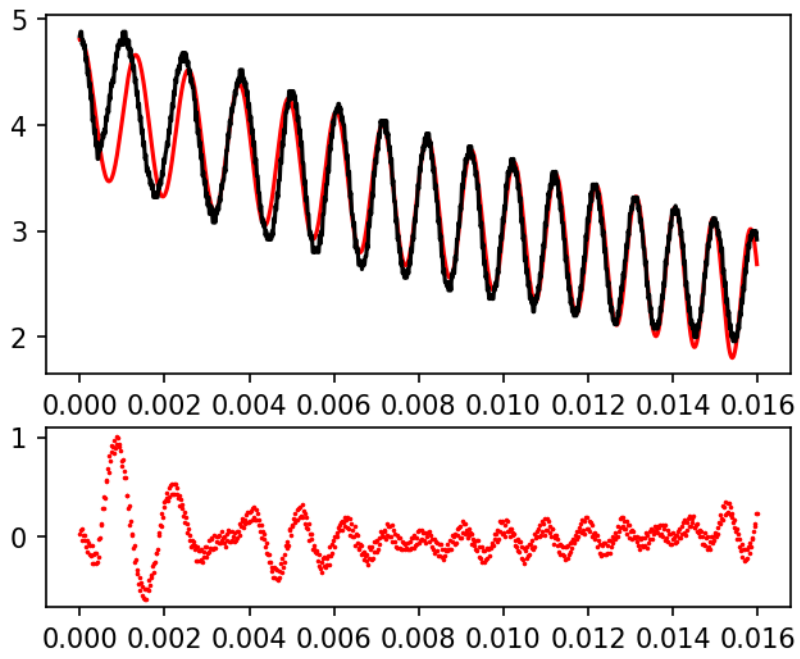
$f_0=1.00\text{e}+14 \pm 4.84\text{e}+06$
 $\alpha=3.32\text{e}+11 \pm 1.21\text{e}+09$
 $A=6.32\text{e}-01 \pm 8.60\text{e}-03$
 $a=3.54\text{e}+00 \pm 1.72\text{e}-02$
 $b=-1.13\text{e}+02 \pm 1.51\text{e}+00$
 $\beta=5.46\text{e}+12 \pm 6.60\text{e}+10$

run15



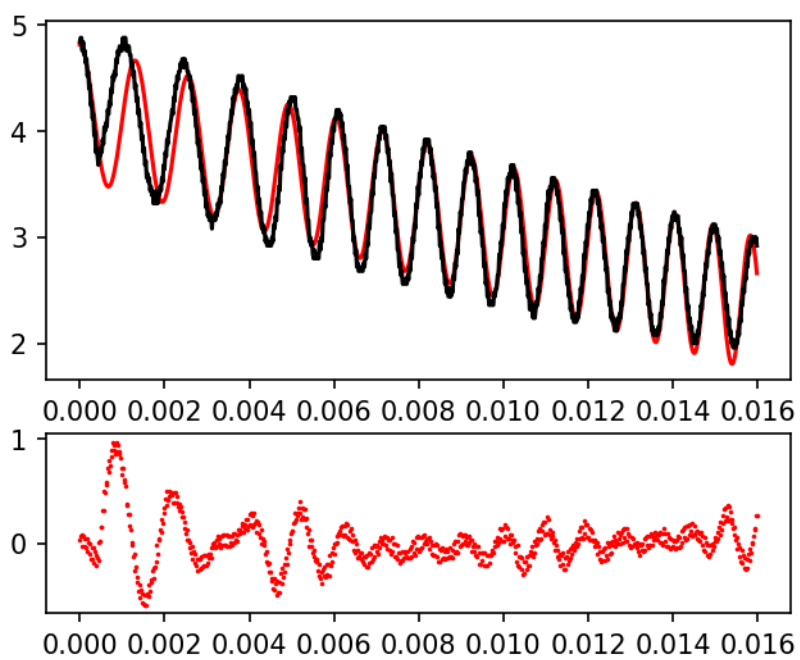
$f_0=1.00\text{e}+14 \pm 4.64\text{e}+06$
 $\alpha=3.33\text{e}+11 \pm 1.16\text{e}+09$
 $A=6.33\text{e}-01 \pm 8.30\text{e}-03$
 $a=3.55\text{e}+00 \pm 1.66\text{e}-02$
 $b=-1.13\text{e}+02 \pm 1.46\text{e}+00$
 $\beta=5.42\text{e}+12 \pm 6.34\text{e}+10$

run16



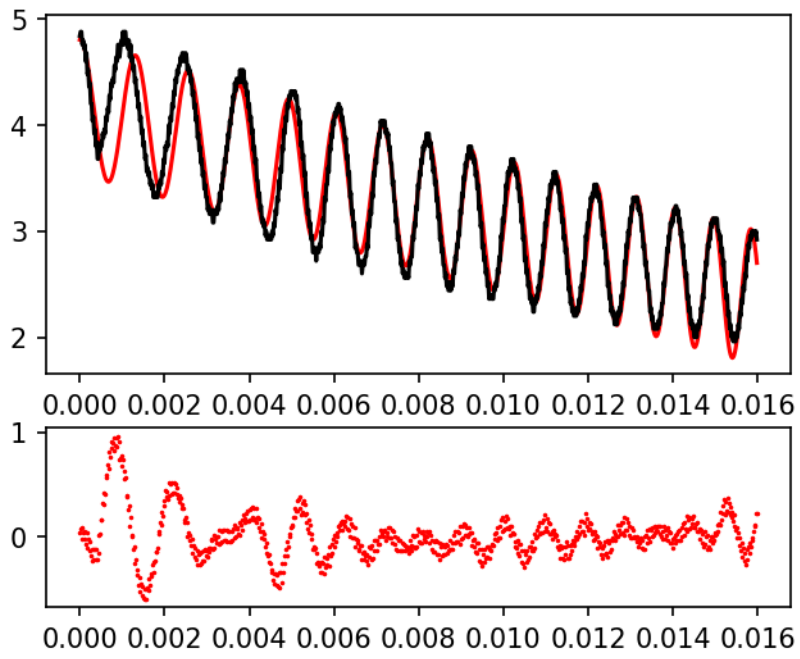
$f_0=1.00\text{e}+14 \pm 4.74\text{e}+06$
 $\alpha=3.32\text{e}+11 \pm 1.18\text{e}+09$
 $A=6.31\text{e}-01 \pm 8.49\text{e}-03$
 $a=3.55\text{e}+00 \pm 1.70\text{e}-02$
 $b=-1.13\text{e}+02 \pm 1.49\text{e}+00$
 $\beta=5.47\text{e}+12 \pm 6.48\text{e}+10$

run17



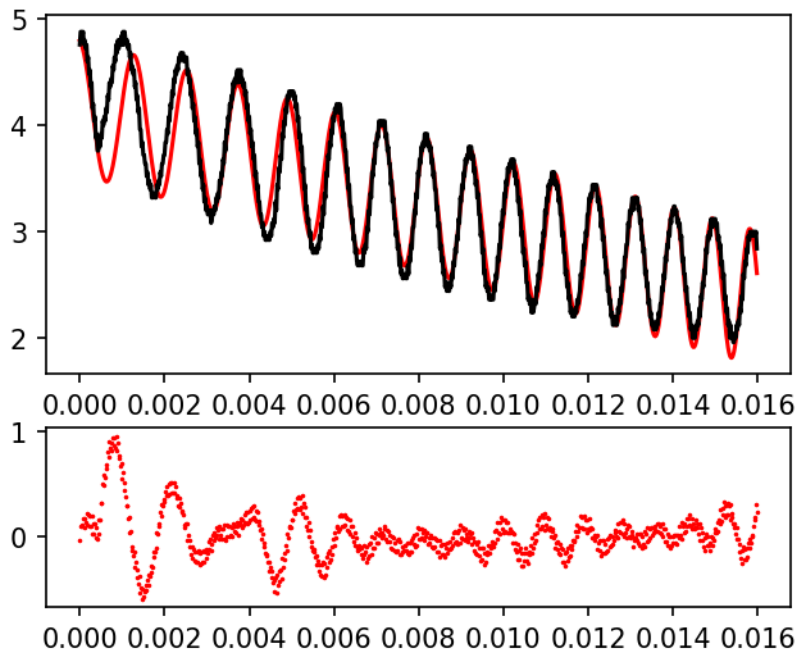
$f_0=1.00\text{e}+14 \pm 4.75\text{e}+06$
 $\alpha=3.31\text{e}+11 \pm 1.19\text{e}+09$
 $A=6.32\text{e}-01 \pm 8.52\text{e}-03$
 $a=3.54\text{e}+00 \pm 1.70\text{e}-02$
 $b=-1.13\text{e}+02 \pm 1.49\text{e}+00$
 $\beta=5.49\text{e}+12 \pm 6.51\text{e}+10$

run18

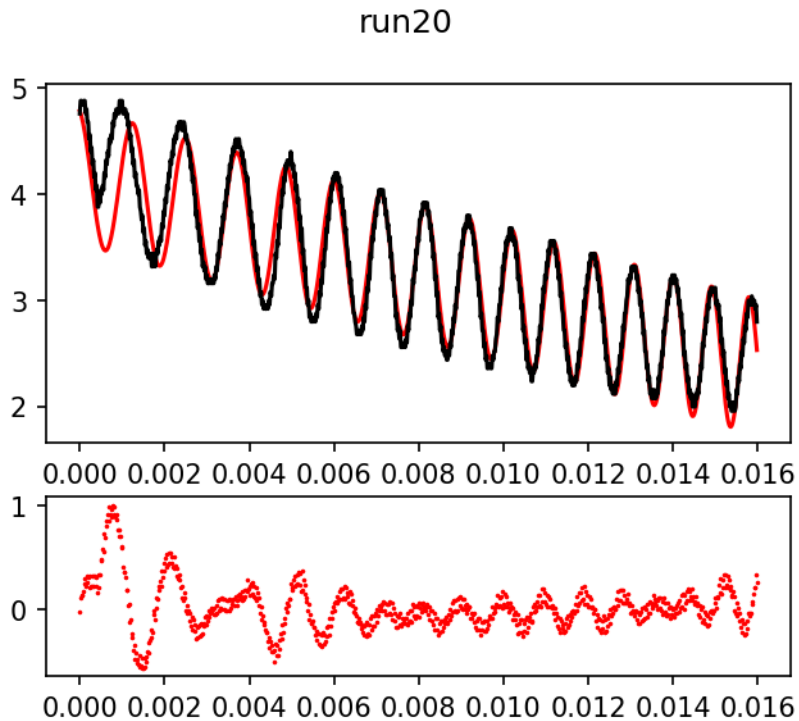


$f_0=1.00\text{e}+14 \pm 4.75\text{e}+06$
 $\alpha=3.31\text{e}+11 \pm 1.19\text{e}+09$
 $A=6.32\text{e}-01 \pm 8.63\text{e}-03$
 $a=3.54\text{e}+00 \pm 1.73\text{e}-02$
 $b=-1.12\text{e}+02 \pm 1.51\text{e}+00$
 $\beta=5.51\text{e}+12 \pm 6.52\text{e}+10$

run19



$f_0 = 1.00\text{e}+14 \pm 4.61\text{e}+06$
 $\alpha = 3.31\text{e}+11 \pm 1.15\text{e}+09$
 $A = 6.36\text{e}-01 \pm 8.48\text{e}-03$
 $a = 3.54\text{e}+00 \pm 1.70\text{e}-02$
 $b = -1.12\text{e}+02 \pm 1.48\text{e}+00$
 $\beta = 5.47\text{e}+12 \pm 6.33\text{e}+10$



Displaying gap results

```
# calculate the average frequency gap
average_freq_gaps = np.mean([hyperfine_freq_gaps['run'+str(i)] for i
in range(1, 20)], axis = 0)
std_freq_gap = np.std([hyperfine_freq_gaps['run'+str(i)] for i in
range(1, 20)], axis=0)
average_ground_gaps = np.mean([ground_freq_gaps['run'+str(i)] for i in
range(1, 20)], axis = 0)
std_ground_gap = np.std([ground_freq_gaps['run'+str(i)] for i in
range(1, 20)], axis=0)

for avg_gap, std_gap in zip(average_freq_gaps, std_freq_gap):
    print(f'Average Frequency Gap: {avg_gap:.2e} ± {std_gap:.2e}')

# plot the average frequency gap
plt.figure(figsize=(10,5), dpi=150)
hyperfine_indices = [0,1,3,4,6,7,9,10]
doppler_indices = [2,5,8,11]

# plotting gaps between doppler broadened peaks
plt.subplot(211)
plt.title('Average Frequency Gaps Between Doppler Broadened Peaks')
plt.errorbar([2, 5, 8], average_freq_gaps[[2,5,8]],
yerr=std_freq_gap[[2,5,8]], fmt='o', capsize=10, label = 'Between
Doppler \nbroadened peaks', color = 'purple')
```

```

plt.xlim(-1, 12)
plt.xticks([0, 1, 3, 4, 6, 7, 9, 10], labels=[])
plt.grid(True, linestyle='--', alpha=0.5)
plt.legend()

# gaps between hyperfine peaks
plt.subplot(212)
plt.title('Average Frequency Gaps Between Hyperfine Peaks')
plt.errorbar([0, 1], average_freq_gaps[[0,1]],
yerr=std_freq_gap[[0,1]], fmt='o', capsize=10, label = r'Within
 $\text{\textsuperscript{87}Rb}$  ( $F=2 \rightarrow F^{\prime}$ )')
plt.errorbar([3, 4], average_freq_gaps[[3,4]],
yerr=std_freq_gap[[3,4]], fmt='o', capsize=10, label = r'Within
 $\text{\textsuperscript{85}Rb}$  ( $F=3 \rightarrow F^{\prime}$ )')
plt.errorbar([6, 7], average_freq_gaps[[6,7]],
yerr=std_freq_gap[[6,7]], fmt='o', capsize=10, label = r'Within
 $\text{\textsuperscript{85}Rb}$  ( $F=2 \rightarrow F^{\prime}$ )')
plt.errorbar([9, 10], average_freq_gaps[[9,10]],
yerr=std_freq_gap[[9,10]], fmt='o', capsize=10, label = r'Within
 $\text{\textsuperscript{87}Rb}$  ( $F=1 \rightarrow F^{\prime}$ )')

plt.xticks([0, 1, 3, 4, 6, 7, 9, 10], ['1-2', '2-3', '2-3', '3-4', '1-
2', '2-3', '0-1', '1-2'])
plt.xlabel(r' $F^{\prime}-F$  Gaps')
plt.legend(ncol = 2)
plt.xlim(-1, 12)
plt.grid(True, linestyle='--', alpha=0.5)

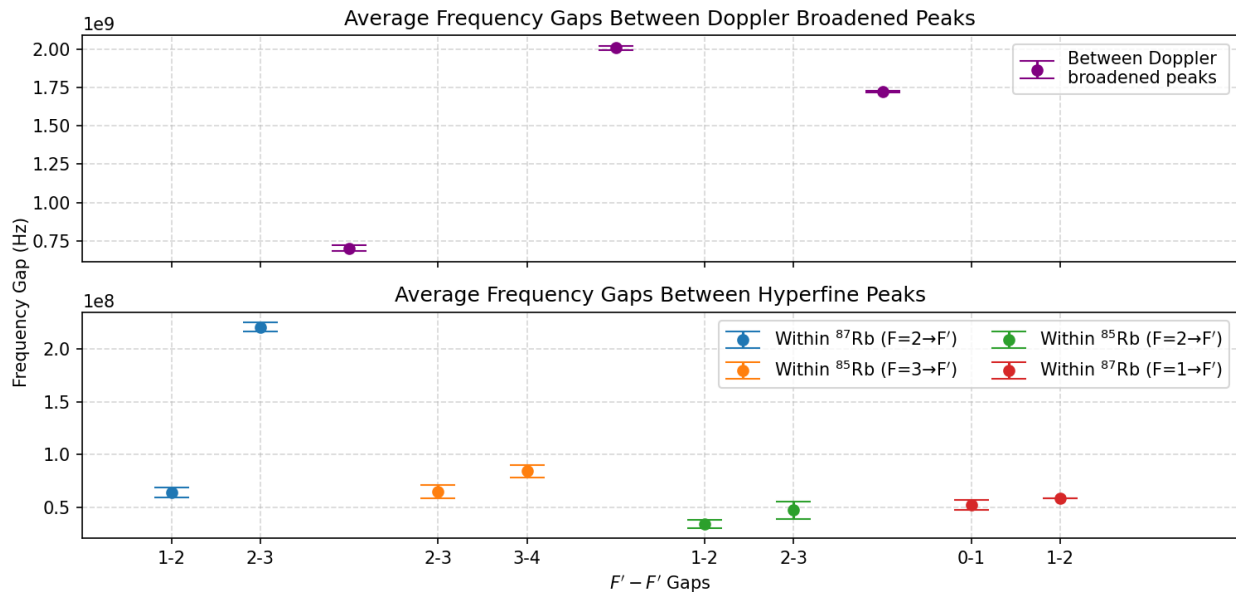
# Add a common y-axis label
plt.gcf().text(0.0, 0.5, 'Frequency Gap (Hz)', va='center',
rotation='vertical')
plt.tight_layout()
plt.show()

```

```

Average Frequency Gap: 6.41e+07 ± 4.61e+06
Average Frequency Gap: 2.21e+08 ± 4.10e+06
Average Frequency Gap: 7.03e+08 ± 1.95e+07
Average Frequency Gap: 6.52e+07 ± 6.15e+06
Average Frequency Gap: 8.45e+07 ± 5.94e+06
Average Frequency Gap: 2.01e+09 ± 1.31e+07
Average Frequency Gap: 3.46e+07 ± 4.01e+06
Average Frequency Gap: 4.73e+07 ± 8.55e+06
Average Frequency Gap: 1.72e+09 ± 4.97e+06
Average Frequency Gap: 5.26e+07 ± 4.78e+06
Average Frequency Gap: 5.83e+07 ± 3.09e+04

```



```
# Ground gaps
print(f'85Rb: {average_ground_gaps[0]*1e-6} ± {std_ground_gap[0]*1e-6}
MHz')
print(f'87Rb: {average_ground_gaps[1]*1e-6} ± {std_ground_gap[1]*1e-6}
MHz')

# Ground gaps
theory_ground = np.array([3.036, 6.835]) * 1e9
chi_square_ground = (theory_ground - average_ground_gaps)*2 /
std_ground_gap*2
print(chi_square_ground)

85Rb: 2173.7390685608552 ± 11.813031989399379 MHz
87Rb: 4997.330907320723 ± 7.732974193318729 MHz
[291.96938846 950.56264084]
```

Multipanel plot: showing data processing on a single run

Showing:

- The pump - no pump data that we used to get the peaks
- The peaks highlighted and annotated
- the interferometer data for that run with fit and residuals

```
probe_data = raw_probe_pump_data['run1']['y2'] -
raw_probe_data['nopump1']['y2']
time = raw_probe_data['nopump1']['x'] / 10

# Plot 1: Raw Data
plt.figure(figsize=(5, 3), dpi=300)
plt.title('a) Raw Data', loc='left')
```

```

plt.plot(time, raw_probe_data['nopump1']['y2'], label='No Pump',
color='k', linewidth=0.3)
plt.plot(time, raw_probe_pump_data['run1']['y2'], label='Pump',
color='r', linewidth=0.3)
plt.legend()
plt.xlim(0.001, 0.015)
plt.ylabel('Voltage (V)')
plt.xlabel('Time (s)')
plt.savefig('figures/raw_data.pdf')
plt.show()

```

Plot 2: Subtracted Data

```

plt.figure(figsize=(5, 3), dpi=300)
plt.title('b) Subtracted Data', loc='left')
plt.plot(time, probe_data, label='Difference', color='b',
linewidth=0.3, zorder=0)
plt.scatter(time[hyperfine_peaks['run1']],
probe_data[hyperfine_peaks['run1']], color='r', s=5, alpha=0.5)
plt.legend()
plt.xlim(0.001, 0.015)
plt.xlabel('Time (s)')
plt.axvspan(time[118], time[157], color='lightblue', alpha=0.5,
zorder=0)
plt.axvspan(time[245], time[265], color='mistyrose', alpha=0.5,
zorder=0)
plt.axvspan(time[503], time[512], color='beige', alpha=0.5, zorder=0)
plt.axvspan(time[697], time[708], color='lavender', alpha=0.5,
zorder=0)
plt.savefig('figures/subtracted_data.pdf')
plt.show()

```

Plot 3: Interferometer

```

plt.figure(figsize=(5, 3), dpi=300)
plt.title('c) Interferometer', loc='left')
plt.plot(time, raw_probe_pump_data['run1']['y3'],
label='Interferometer', color='k', linewidth=0.7, zorder=0)
plt.plot(time, interference_model(time, *parameters), label='Fit',
alpha=0.7, color='purple')
plt.legend()
plt.xlim(0.001, 0.015)
plt.xticks([])
plt.savefig('figures/interferometer.pdf')
plt.show()

```

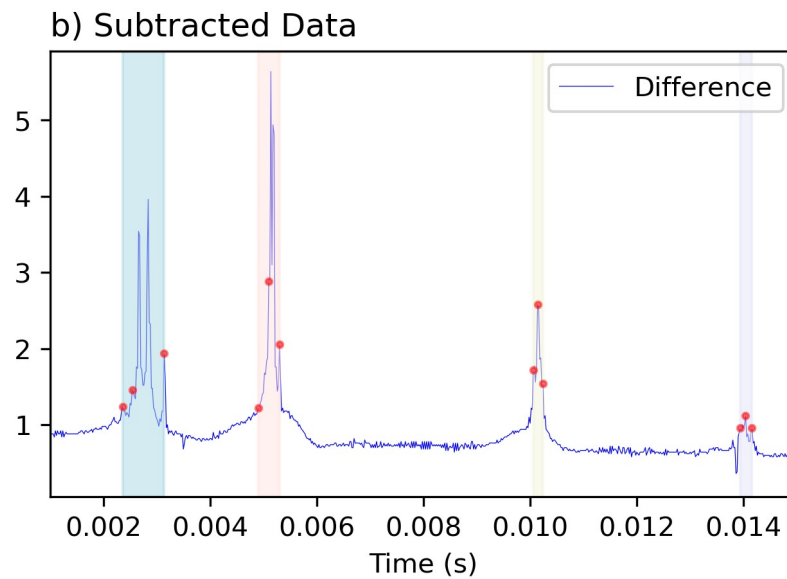
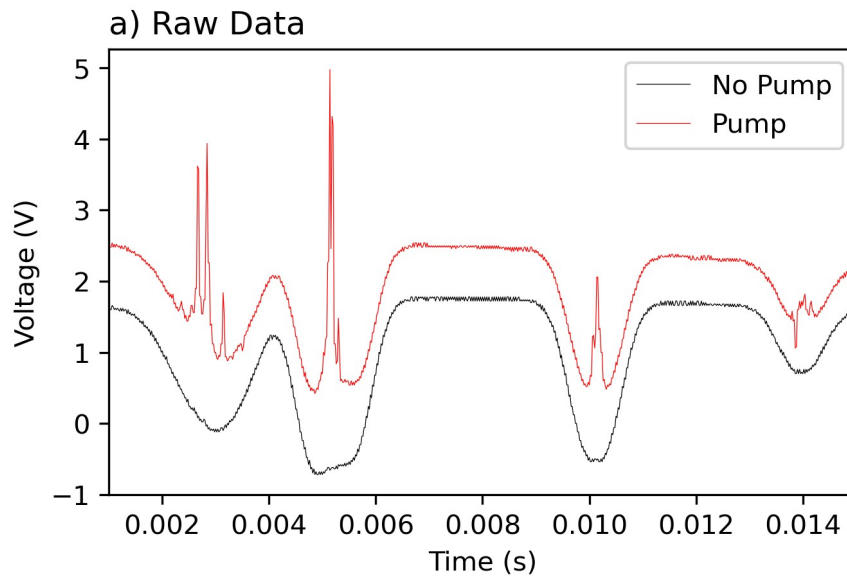
Plot 4: Residuals

```

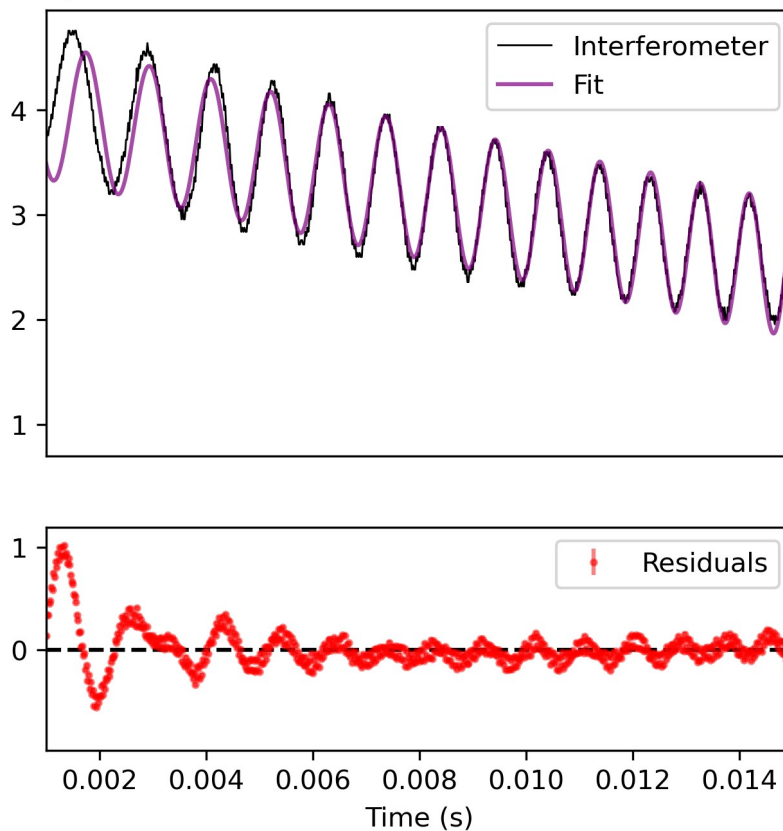
plt.figure(figsize=(5, 1.5), dpi=300)
plt.errorbar(time, raw_probe_pump_data['run1']['y3'] -
interference_model(time, *parameters), yerr=0.02, label='Residuals',
color='r', alpha=0.5, fmt='o', markersize=2)
plt.hlines(0, 0.001, 0.015, color='k', linestyle='--')

```

```
plt.xlim(0.001, 0.015)
plt.ylim(-0.98, 1.2)
plt.legend()
plt.xlabel('Time (s)')
plt.savefig('figures/residuals.pdf')
plt.show()
```



c) Interferometer



```
t1 = time[115:165]
t2 = time[240:270]
t3 = time[495:520]
t4 = time[685:720]

p1 = probe_data[115:165]
p2 = probe_data[240:270]
p3 = probe_data[495:520]
p4 = probe_data[685:720]

# make a subplot for each region
fig, axs = plt.subplots(1, 4, figsize=(8,4), dpi=300)
fig.suptitle('d) Zoomed in Regions of the Data')
fig.tight_layout(rect=[0, 0.03, 1, 0.95])
fig.supxlabel('Time (s)')

# plot the data and peaks from that region
axs[0].set_title(r'$\gamma$Rb (F=2$\rightarrow$F'$\rightarrow$F'))
axs[0].plot(t1, p1, color='b', linewidth=0.7, zorder=0)
axs[0].scatter(t1[[118 - 115, 127 - 115, 157 - 115]], p1[[118 - 115, 127 - 115, 157 - 115]], color='r', s = 7, label = 'Resonance peaks')
axs[0].set_ylabel('Voltage (V)')
axs[0].set_facecolor('lightblue')
```

```

axs[1].set_title(r'$^{85}$Rb (F=3$\rightarrow$F$^{\prime}$)')
axs[1].plot(t2, p2, color='b', linewidth=0.7, zorder=0)
axs[1].scatter(t2[[245-240, 255-240, 265-240]], p2[[245-240, 255-240, 265-240]], color='r', s = 7, label = 'Resonance peaks')
axs[1].set_facecolor('mistyrose')

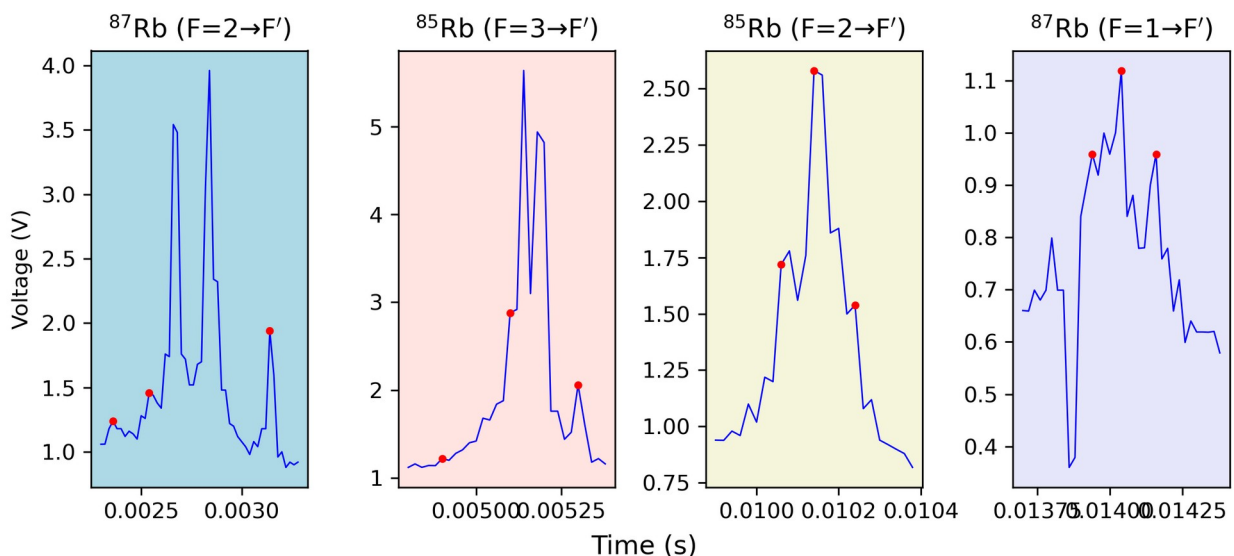
axs[2].set_title(r'$^{85}$Rb (F=2$\rightarrow$F$^{\prime}$)')
axs[2].plot(t3, p3, color='b', linewidth=0.7, zorder=0)
axs[2].scatter(t3[[503-495, 507-495, 512-495]], p3[[503-495, 507-495, 512-495]], color='r', s = 7, label = 'Resonance peaks')
axs[2].set_facecolor('beige')

axs[3].set_title(r'$^{87}$Rb (F=1$\rightarrow$F$^{\prime}$)')
axs[3].plot(t4, p4, color='b', linewidth=0.7, zorder=0)
axs[3].scatter(t4[[697-685, 702-685, 708-685]], p4[[697-685, 702-685, 708-685]], color='r', s = 7, label = 'Resonance peaks')
axs[3].set_facecolor('lavender')

plt.savefig('figures/zoomins.pdf')
# plt.tight_layout()

```

d) Zoomed in Regions of the Data



Chi-square calculations

```

# chi square = observed - expected / error

# chi square for each run
theory = np.array([0.157, 0.287, 100, 0.063, 0.121, 100, 0.029, 0.663,
100, 0.072, 0.157]) * 1e9

```

```

# get chi square for 1st value by taking 1st gap from each run
chi_square = (theory - average_freq_gaps[0])**2 / std_freq_gap[0]**2
# chi_square = np.sum(chi_square)
# print(f'Chi Square: {chi_square:.2f}')
chi_squares = []

for i in range(11):
    chi_squares.append((theory[i] - average_freq_gaps[i])**2 /
std_freq_gap[i]**2)

print(np.array(chi_squares)[[0,1,3,4,6,7,9,10]])

[4.06154042e+02 2.61230218e+02 1.25655282e-01 3.77164985e+01
 1.97806962e+00 5.18753359e+03 1.65616768e+01 1.02275097e+07]

```

Fitting FWHM of no-pump runs

```

def gaussian(x, a, b, c, d):
    return a * np.exp(-(x - b)**2 / (2 * c**2)) + d

data_time = raw_probe_data['nopump1']['x'] / 10 # convert to seconds
data_voltage = raw_probe_data['nopump1']['y2']

colors = ['blue', 'red', 'maroon', 'indigo']
fill_colors = ['lightblue', 'mistyrose', 'beige', 'lavender']
fit_labels = [r'$^{87}$Rb (F=2$\to$F$^{\prime}$)', r'$^{85}$Rb (F=3$\to$F$^{\prime}$)', r'$^{85}$Rb (F=2$\to$F$^{\prime}$)', r'$^{87}$Rb (F=1$\to$F$^{\prime}$)']

slice_indices = [80, 200, 212, 330, 470, 540, 665, 730]
slice_edges = data_time[slice_indices]

# Figure
plt.figure(figsize=(5,3), dpi=250)

# Plot data
plt.scatter(data_time, data_voltage, color='k', s=0.1, label = 'Data')

# Fit the data
fit_dict = {}
for i,(start,stop) in enumerate(zip(slice_indices[:-1:2],
slice_indices[1::2])):
    plt.axvspan(data_time[start], data_time[stop],
color=fill_colors[i], zorder = 0)

    # initial_guess = [-1, 1e-2, 1e-4, -1]
    popt, pcov = opt.curve_fit(gaussian, data_time[start:stop],
data_voltage[start:stop],
                                p0=[-1, data_time[int((stop - start) /
2) + start], 0.0002, -1], maxfev = 100000)

```



```

# Add to dictionary
fit_dict[f'popt_{i}'] = popt
fit_dict[f'err_{i}'] = np.sqrt(np.diag(pcov))
# Plot the fit
fit_voltage = gaussian(data_time[start:stop], *popt)
print(popt)

plt.plot(data_time[start:stop], fit_voltage, label=fit_labels[i],
color=colors[i])

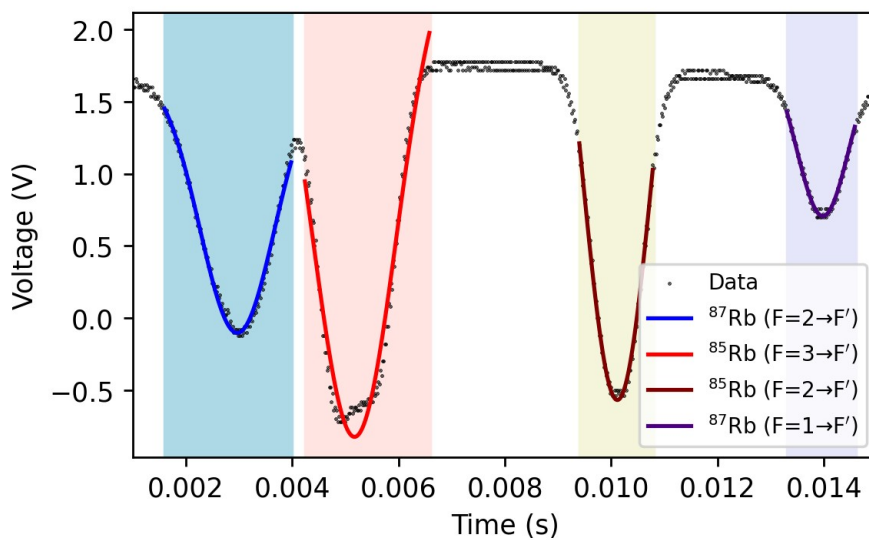
plt.xlim(0.001, 0.015)

# Labels
plt.xlabel('Time (s)')
plt.ylabel('Voltage (V)')
plt.legend(fontsize=8)

[ -1.83043700e+00  2.96805416e-03  7.01730154e-04  1.72707776e+00]
[ -3.50869244e+00  5.17254915e-03  7.86295755e-04  2.68452795e+00]
[ -4.08912749e+00  1.01148937e-02  6.68574739e-04  3.52049988e+00]
[ -1.18513236e+00  1.39790464e-02  4.93950693e-04  1.89234957e+00]

<matplotlib.legend.Legend at 0x7f8f50a54400>

```



```

plt.figure(figsize=(6,2.5), dpi=200)

# Plot residuals for the fits
for i,(start,stop) in enumerate(zip(slice_indices[:-1:2],
slice_indices[1::2])):
    # Get data
    data = data_voltage[start:stop]

```

```

time = data_time[start:stop]
# Get fit voltage
fit_voltage = gaussian(time, *fit_dict[f'popt_{i}'])
# Get indices where the data isn't 0
nonzero_indices = np.where(data != 0)
# Calculate % residuals
residuals = data[nonzero_indices] - fit_voltage[nonzero_indices]
plt.scatter(time[nonzero_indices], residuals, label=fit_labels[i],
s=2, color=colors[i])

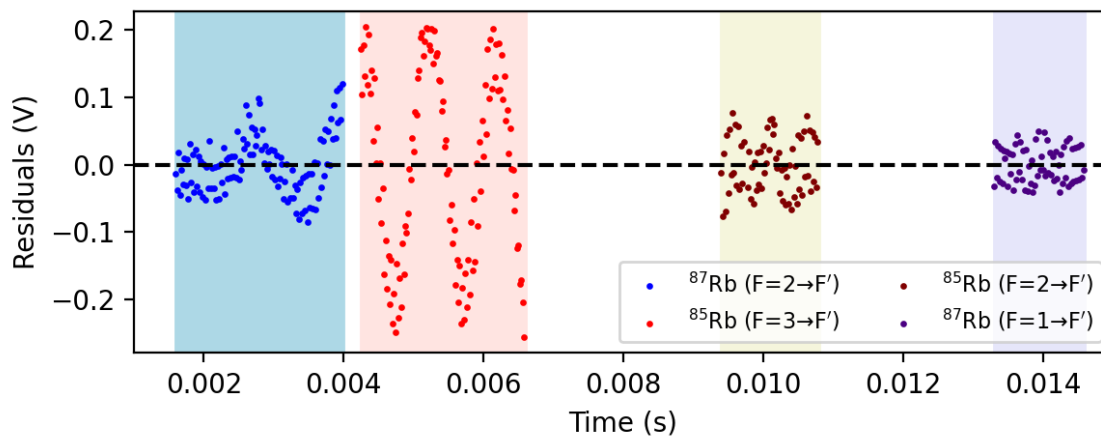
plt.axvspan(data_time[start], data_time[stop],
color=fill_colors[i], zorder = 0)

# x axis, in seconds
# start_time = data_time[slice_indices[0]]*time_multiplier
# stop_time = data_time[slice_indices[-1]]*time_multiplier
plt.xlim(0.001, 0.015)
plt.hlines(0, 0.001, 0.015, color='k', linestyle='--')

# Labels
plt.xlabel('Time (s)')
plt.ylabel('Residuals (V)')
# plt.ylim(-50,50)
plt.legend(loc='lower right', fontsize=8, ncol = 2)

plt.tight_layout()

```



```

# getting FWHM of all fits
def fwhm_calc(popt):
    return np.sqrt(4*np.log(2)) * popt[2]

# FWHM for each fit
fwhm = []
fwhm_err = []

```

```

for i in range(4):
    fwhm.append(fwhm_calc(fit_dict[f'popt_{i}']))
    fwhm_err.append(fwhm_calc(fit_dict[f'err_{i}']))
# print for each curve
for i in range(int(len(fit_dict)/2)):
    print(f'FWHM for {fit_labels[i]}: {fwhm[i] * 1e3:.2e} ±
{fwhm_err[i] * 1e3:.2e} ms')

# chi square of fits
chi_square = [np.sum((data_voltage[start:stop] -
gaussian(data_time[start:stop], *fit_dict[f'popt_{i}']))*2 /
gaussian(data_time[start:stop], *fit_dict[f'popt_{i}'])))
                for i, (start, stop) in enumerate(zip(slice_indices[:-
1:2], slice_indices[1::2]))]

print(chi_square)

FWHM for  $\text{Rb}$  ( $F=2\hbar$ ): 1.17e+00 ± 2.37e-02 ms
FWHM for  $\text{Rb}$  ( $F=3\hbar$ ): 1.31e+00 ± 5.50e-02 ms
FWHM for  $\text{Rb}$  ( $F=2\hbar$ ): 1.11e+00 ± 7.57e-02 ms
FWHM for  $\text{Rb}$  ( $F=1\hbar$ ): 8.22e-01 ± 5.76e-02 ms
[-0.18072124771471762, -4.1540185040474364, -1.559666290633998,
0.04971367211340433]

# converting the FWHM to frequency space using frequency model and
parameters saved for run 1
def to_freq(t, f0, alpha, beta):
    return f0 + alpha * t + beta*t**2

fwhm_freqs = to_freq(np.array(fwhm), parameters[0], parameters[1],
parameters[5])
fwhm_freqs_err = to_freq(np.array(fwhm_err), parameters[0],
parameters[1], parameters[5])

for i in range(4):
    print(f'FWHM for {fit_labels[i]}: {fwhm_freqs[i]:.7e} ±
{fwhm_freqs_err[i]:.7e} Hz')

FWHM for  $\text{Rb}$  ( $F=2\hbar$ ): 1.0000082e+14 ±
1.0000042e+14 Hz
FWHM for  $\text{Rb}$  ( $F=3\hbar$ ): 1.0000087e+14 ±
1.0000043e+14 Hz
FWHM for  $\text{Rb}$  ( $F=2\hbar$ ): 1.0000080e+14 ±
1.0000043e+14 Hz
FWHM for  $\text{Rb}$  ( $F=1\hbar$ ): 1.0000070e+14 ±
1.0000043e+14 Hz

# Constants
k_B = 1.380649e-23 # Boltzmann constant in J/K
T_Celsius = 45.6 # Temperature in Celsius
T_Kelvin = T_Celsius + 273.15 # Convert to Kelvin

```

```

mass_85Rb = 1.409993199e-25 # Mass of 85Rb in kg
mass_87Rb = 1.44316060e-25 # Mass of 87Rb in kg

# Convert Celsius to Kelvin
T = T_Kelvin

# Doppler broadening FWHM formula for velocity
def doppler_fwhm(T, mass):
    return np.sqrt(4 * k_B * T * np.log(2) / mass) / 780e-9 #gives
    unitless, m/s divided by m/s

# Calculate FWHM for each isotope, and divide by the central frequency
of the laser
fwhm_85Rb = doppler_fwhm(T_Kelvin, mass_85Rb) / 1e6
fwhm_87Rb = doppler_fwhm(T_Kelvin, mass_87Rb) / 1e6

print(f'FWHM for 85Rb: {fwhm_85Rb} MHz')
print(f'FWHM for 87Rb: {fwhm_87Rb} MHz')

FWHM for 85Rb: 377.14326161450185 MHz
FWHM for 87Rb: 372.78422808562476 MHz

```