Lorentzian Final

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- 1 Objective: Given a Gold nanocube, obtain its wavelength (energy) scans 10 times consecutively via AOTF-VIS. Following this:
- 1.0.1 1. Plot the 10 consecutive scans on a single graph, taking wavelength (nm) in the x-axis and corrected extinction (a.u) in the y-axis.
- 1.0.2 2. Extract the regions in all the scans where a local peak is observed. Plot them together in the same manner as above, using energy (eV) in the x-axis instead.
- 1.0.3 3. Obtain the best-fit Lorentzian curve for all the scans. Extract the respective Full Width at Half Maximum (FWHM) and peak Resonant Energy (E_{res}) for all of the scans. Plot all the best-fits together.
- 1.0.4 4. Plot all the best-fits together but ensure that the curves are normalised w.r.t. peak corrected extinction.
- 1.0.5 5. Obtain Δ FWHM i.e. the FWHM of all scans relative to that of the 1st scan. Plot these values w.r.t. E_{res} .
- 1.0.6 6. Obtain ΔE_{res} i.e. the peak Resonant Energy of all the scans relative to that of the 1st scan. Plot these values w.r.t. E_{res} .
- 1.1 Relevant packages and modules required

```
[1]: import pandas as pd
  import numpy as np
  import matplotlib.pyplot as plt
  import scipy.constants as sc
  import os
  from pylab import cm
  from lmfit.models import LorentzianModel
  from tabulate import tabulate
```

1.2 Function to extract our parameters of interest as well as the number of repeats for the respective scans from the data files

1.3 Function to generate extinction arrays for the given csv file

```
[3]: def extinction(readfile, repeats):
    extinctions = readfile.loc[:,"R(V)"]/readfile.loc[:,"M(V)"]
    extinctions = extinctions.groupby(np.arange(len(extinctions))//repeats[0]).
    →mean()
    return extinctions
```

1.4 Function to store extinction data for the consecutive scans as a 2D array i.e. Scan 1 data will be in column 1, Scan 2 in column 2, and so on

```
[4]: def ext_matrix(extinctions, repeats):
    n_rows = len(extinctions)//repeats[1]
    n_cols = repeats[1]
    extin_matrix=np.zeros((n_rows, n_cols))
    for i in range(n_cols):
        loc = n_rows*i
        extin_matrix[:,i] = extinctions[loc:(loc+n_rows)]
    return extin_matrix
```

1.5 Function that finds the corrected extinctions for all the scans by utilising the reference data

```
[5]: def corr_ext(extin_matrix, extin_ref):
    n_rows, n_cols = extin_matrix.shape
    corr_extin_matrix = np.zeros((n_rows, n_cols))
    for i in range(n_cols):
        corr_extin_matrix[:,i] = 1.00 - (extin_matrix[:,i]/extin_ref)
    return corr_extin_matrix
```

1.6 Function that prepares the extinction 2D array as well as the wavelengths for subsequent plotting as per Objective 1

```
[6]: def data_prep_ob1(wavelengths, corr_extin_matrix):
    n_rows, n_cols = corr_extin_matrix.shape
    xData = np.zeros((n_cols, n_rows))
    yData = np.zeros((n_cols, n_rows))
    for i in range(n_cols):
        xData[i,:] = wavelengths
        yData[i,:] = corr_extin_matrix[:,i]
    return xData, yData
```

1.7 Function that yields the plot as per Objective 1 and 2

```
[7]: def plot_ob1_2(xData = [], yData = [], xLabel = '', yLabel = 'Extinction (a.u.

→)', Title = '', f = ''):
    n_rows, n_cols = yData.shape
    colors = cm.get_cmap('viridis', n_rows)
    plt.figure(num = None, figsize = (10, 7), facecolor = 'w', edgecolor = 'k')
    for i in range(n_rows):
        plt.plot(xData[i,:], yData[i,:], label = "Scan "+str(i+1), color = "
    →colors(i))
        plt.xlabel(xLabel)
        plt.ylabel(yLabel)
        plt.legend(loc='upper right')
        plt.title(Title)
        plt.legend(bbox_to_anchor=(0.99, 0.99), borderaxespad=0)
        plt.show
        plt.savefig(f)
```

1.8 Function that converts the wavelength (in nm) to energy (in eV)

```
[8]: def lambda_to_E(wavelengths):
    return (sc.h*sc.c*10**9/sc.e)/(wavelengths)
```

1.9 Function that yields the wavelength slice of the expected peak as well as the corresponding extinction data slice

```
[9]: def data_slice(xData, yData, left_thres = 650.0, right_thres = 755.0):
    n_rows, n_cols = yData.shape
    xSlice = np.zeros((n_rows, n_cols))
    ySlice = np.zeros((n_rows, n_cols))
    for i in range(n_rows):
        begin_index = np.nonzero(xData[i,:] == left_thres)[0][0]
        end_index = np.nonzero(xData[i,:] == right_thres)[0][0]
        fit_range = slice(begin_index, end_index+1)
```

1.10 Function that obtains the best-fit Lorentzian curves for the consecutive data slices provided

```
def lorentzian_fit(xSlice, ySlice):
    n_rows, n_cols = ySlice.shape
    model = LorentzianModel()
    result_params = []
    result_bestfit = np.zeros((n_rows, n_cols))
    for i in range(n_rows):
        params = model.guess(ySlice[i,:], xSlice[i,:])
        results = model.fit(ySlice[i,:], params, x=xSlice[i,:])
        result_params.append(list(results.params.items()))
        result_bestfit[i,:] = results.best_fit
    return result_params, result_bestfit
```

1.11 Function that extracts the model best-fit parameters held in a dictionary and stores it in a 2D array

```
[11]: def param_dict_array(params):
    n_rows = len(params)
    no_of_params = len(params[0])
    param_set = np.zeros((n_rows, no_of_params))
    for i in range(n_rows):
        for j in range(no_of_params):
            param_set[i,j] = params[i][j][1].value
    return param_set
```

1.12 Function that extracts the FWHM, E_{res} , and height of peak maxima for the 10 consecutive scans

```
[12]: def param_extract(param_set):
    E_res = param_set[:,1]
    fwhm = param_set[:,3]
    height_max_peak = param_set[:,4]
    return fwhm, E_res, height_max_peak
```

1.13 Function that tabulates the FWHM, Δ FWHM, E_{res} , ΔE_{res} , and Height of peak maxima for the 10 consecutive scans

1.14 Function that normalizes the extinction dataset as well as the best-fit extinction dataset w.r.t. height of peak maxima for the 10 consecutive scans as per Objective 4

1.15 Function that plots the best-fit Lorentzian curves along with the corresponding datasets (as dots) for the 10 consecutive scans as per Objective 3 and 4

```
plt.legend(bbox_to_anchor=(0.99, 0.99), borderaxespad=0)
plt.show
plt.savefig(f)
```

1.16 Function that finds Δ FWHM and ΔE_{res} as per Objective 5 and 6

```
[16]: def delta_param(fwhm, E_res):
    n_rows = len(fwhm)
    delta_fwhm = []
    delta_E_res = []
    for i in range(n_rows):
        delta_fwhm.append((fwhm[i] - fwhm[0]))
        delta_E_res.append((E_res[i] - E_res[0]))
    return delta_fwhm, delta_E_res
```

1.17 Function that plots Δ FWHM (meV) and ΔE_{res} (meV) versus E_{res} (eV) for the 10 consecutive scans as per Objective 5 and 6

1.18 Function that tabulates the Δ FWHM (meV) and ΔE_{res} (meV) for the 10th scan

```
[18]: def scan10_table(delta_fwhm, delta_E_res, f):
    table = [['∆FWHM_cube (meV)', '∆E_res_cube (meV)'], [delta_fwhm[-1]*1000,
    →delta_E_res[-1]*1000]]
    print(tabulate(table, headers='firstrow', tablefmt='fancy_grid', floatfmt=".
    →4f"))
    table_df=pd.DataFrame(table)
    table_df.to_csv(f, index=False, header=False)
```

1.19 Main function that executes everything required from start to finish

```
[19]: def Nanocube Lorentzian(f1 = "10ConsScan cube9.csv", f2 = "10ConsScan cube9ref.
      ⇔csv"):
         cwd = os.getcwd()
         csv_dir = '/csv/'
         fig_dir = '/Figures/'
         hits, repeats1 = parameter_formatting WL(cwd + csv_dir + f1, 86, 4, 3)
         miss, repeats2 = parameter formatting WL(cwd + csv dir + f2, 84, 3, 2)
         hits_ext = extinction(hits, repeats1)
         miss_ext = extinction(miss, repeats2)
         hits_ext_matrix = ext_matrix(hits_ext, repeats1)
         corr_hits_ext_matrix = corr_ext(hits_ext_matrix, miss_ext)
         wavelengths = miss.loc[::repeats2[0],'Wavelength(nm)'].values
         xdata, ydata = data_prep_ob1(wavelengths, corr_hits_ext_matrix)
         plot_ob1_2(xdata, ydata, 'Wavelength (nm)', 'Extinction (a.u.)', __
      → 'Consecutive Wavelength Scans on Nanocube', cwd + fig_dir +
      xslice1, yslice = data_slice(xdata, ydata)
         xslice = lambda_to_E(xslice1)
         plot_ob1_2(xslice, yslice, 'Energy (eV)', 'Extinction (a.u.)', 'Consecutive<sub>□</sub>
      →Wavelength Scans on Nanocube', cwd + fig_dir + 'Sliced_energy_scan.png')
         fit_params, fit_ydata = lorentzian_fit(xslice, yslice)
         fit_param_set = param_dict_array(fit_params)
         FWHM, E_res, height_max_peak = param_extract(fit_param_set)
         plot_ob3_4(xslice, yslice, fit_ydata, 'Energy (eV)', 'Extinction (a.u.)', u
      →'Best-fit Lorentzian Curves for Consecutive Scans', cwd + fig_dir + L
      → 'Best-fit_Lorentzian_with_dataset.png')
         plot_ob1_2(xslice, fit_ydata, 'Energy (eV)', 'Extinction (a.u.)', 'Best-fit_u
      →Lorentzian Curves for Consecutive Scans', cwd + fig_dir +
      yslice_norm, fit_ydata_norm = extinction_normalize(yslice, fit_ydata,_u
       →height_max_peak)
```

```
plot_ob3_4(xslice, yslice_norm, fit_ydata_norm, 'Energy (eV)', 'Extinction_

(a.u.)', 'Normalized Best-fit Lorentzian Curves for Consecutive Scans', cwd_

+ fig_dir + 'Best-fit_Lorentzian_with_dataset_normalized.png')

plot_ob1_2(xslice, fit_ydata_norm, 'Energy (eV)', 'Extinction (a.u.)',__

'Normalized Best-fit Lorentzian Curves for Consecutive Scans', cwd + fig_dir_

+ 'Best-fit_Lorentzian_normalized.png')

delta_FWHM, delta_E_res = delta_param(FWHM, E_res)

plot_ob5_6(E_res, delta_FWHM, 'E_res (eV)', 'AFWHM (meV)', 'Change in FWHM__

+ vs E_res', cwd + fig_dir + 'Delta_FWHM_vs_E-res.png')

plot_ob5_6(E_res, delta_E_res, 'E_res (eV)', 'AE_res (meV)', 'Change in__

+ E_res vs E_res', cwd + fig_dir + 'Delta_E-res_vs_E-res.png', 'lower right')

param_table(FWHM, delta_FWHM, E_res, delta_E_res, height_max_peak, cwd +__

+ csv_dir + 'Parameters_of_interest.csv')

scan10_table(delta_FWHM, delta_E_res, cwd + csv_dir + 'Cube_parameters.csv')
```

1.20 Just call the main function!

[20]: Nanocube_Lorentzian()

Scan no. Height of pea	FWHM (eV) k maxima (a.u.)	ΔFWHM (meV)	E_res (eV)	ΔE_res (meV)
Scan 1 0.0327	0.1299	0.0000	1.7328	0.0000
Scan 2 0.0241	0.1724	42.4778	1.7324	-0.3949
Scan 3 0.0338	0.1463	16.3927	1.7387	5.9443
Scan 4 0.0313	0.1463	16.3442	1.7507	17.9328
Scan 5 0.0363	0.1417	11.7528	1.7500	17.2166

Scan 6 0.0319	0.1517	21.7167	1.7614	28.6154
Scan 7 0.0299	0.1329	2.9923	1.7673	34.4895
Scan 8 0.0388	0.1394	9.5031	1.7784	45.5542
Scan 9 0.0347	0.1531	23.2088	1.7974	64.5580
Scan 10 0.0403	0.1372	7.2526	1.8071	74.2891

 Δ FWHM_cube (meV) Δ E_res_cube (meV) 7.2526 74.2891















