

visualization_demo

May 25, 2022

```
[1]: import sys
import numpy as np
import matplotlib.pyplot as plt
from matplotlib import collections as mc

import os
import time
import importlib as imp

import flow_fields as flow
import autocorrelation as ac
import spherical_harmonics as sh
for mod in [flow, ac, sh]:
    imp.reload(mod) # Updates any customizations you've made to the module

# Set matplotlib format
SMALL_SIZE = 12
MEDIUM_SIZE = 15
BIGGER_SIZE = 18

plt.rc('font', size=SMALL_SIZE) # controls default text sizes
plt.rc('axes', titlesize=SMALL_SIZE) # fontsize of the axes title
plt.rc('axes', labelsiz= MEDIUM_SIZE) # fontsize of the x and y labels
plt.rc('xtick', labelsiz=SMALL_SIZE) # fontsize of the tick labels
plt.rc('ytick', labelsiz=SMALL_SIZE) # fontsize of the tick labels
plt.rc('legend', fontsize=SMALL_SIZE) # legend fontsize
plt.rc('figure', titlesize=BIGGER_SIZE) # fontsize of the figure title
```

1 Load Data and Metadata

```
[2]: # Load preprocessed 4D-STEM data and manually add metadata
path = '../data/4DSTEM/'
filename = 'sample_preprocessed_data_80x80_ss=10_cl=480__q 1.00 3.00 0.05 a 5.
→00.npy'

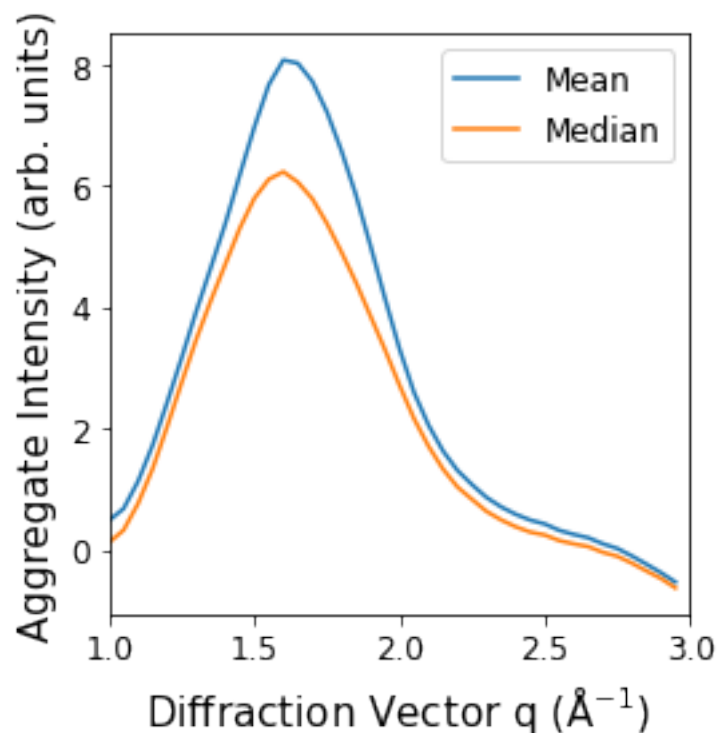
q_min, q_max, q_step = (1.0, 3.0, 0.05)
```

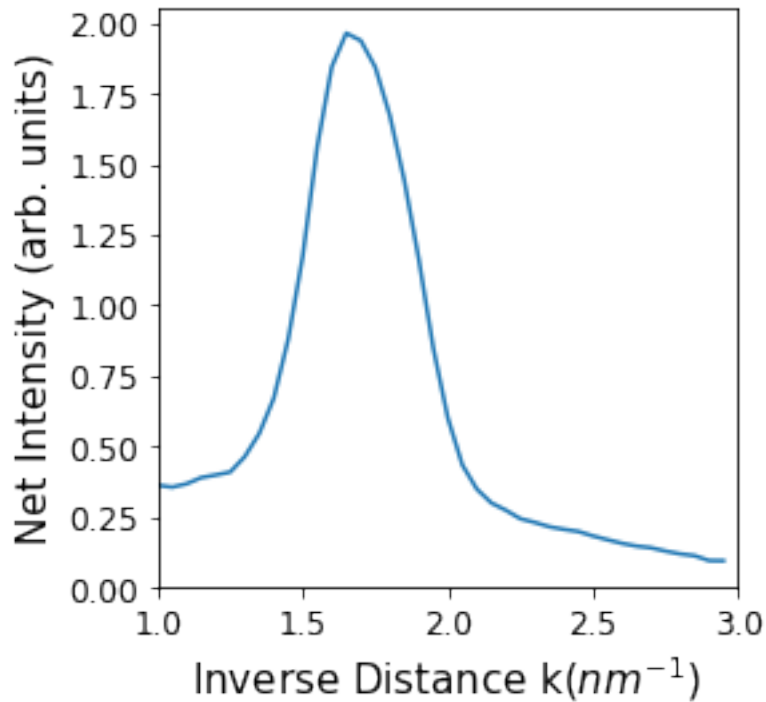
```
angle_step = 5
step_size = 10 # nm. This is the distance separating each diffraction pattern,
↳ in the 4D-STEM scan.

data = np.load(path + filename)
```

2 Choose Integration Range

```
[3]: # Preview Peak Positions
aggregate_plot, difference_plot = flow.preview_intensity_vs_q(data, q_min,
↳ q_max, q_step)
```





```
[4]: # Use the peak preview to choose the integration range for your peaks.

# Note that there are two convoluted peaks represented in the preview, and we
    ↳ only want one of them.

peak_start = 1.6
peak_end = 1.8 # Inclusive

integrated_intensity = flow.integrate_peaks(data, q_min, q_max, q_step,
    ↳ peak_start, peak_end)
prepped_intensity = flow.format_flow_inputs(integrated_intensity, rotate=True)
```

3 Create Flow Plot

```
[5]: peak_width = 40 # degrees
peaks_per_site = 1.2 # This is manually set by the user by viewing the peaks
    ↳ present in the diffraction patterns.

peak_positions = flow.select_peaks(prepped_intensity, peak_width,
    ↳ peaks_per_site)
```

Finding Peaks...

Cutoff = 1.06

Peak Width = 40

7680 total peaks, which is an average of 1.20 peaks per grid square

```
[6]: # Create line seeds at each peak
seed_density = 2
line_seeds = flow.seed_lines(peak_positions, step_size,
    ↳seed_density=seed_density)

# Extend line seeds to create full lines
bend_tolerance = 20
curve_resolution = 2
propagated_lines = flow.propagate_lines(line_seeds, peak_positions, step_size,
    ↳bend_tolerance,
                                     curve_resolution=curve_resolution,
    ↳max_grid_length=100)

# Show a preview, using a subset of the propagated lines
preview_sparsity = 20
propagated_image = flow.plot_solid_lines(propagated_lines, min_length=2,
    ↳sparsity=preview_sparsity)
plt.show()

# Thin out lines, reducing overlap between lines and creating a more
    ↳homogeneous line density. This prevents the
# illusion of high density in regions with good alignment, and makes the image
    ↳more readable.
line_spacing = 1
spacing_resolution = 10
angle_spacing_degrees = 10
max_overlap_fraction = 0.5
trimmed_lines = flow.trim_lines(propagated_lines, prepped_intensity.shape,
    ↳step_size,
                                     line_spacing, spacing_resolution,
    ↳angle_spacing_degrees,
                                     max_overlap_fraction=max_overlap_fraction,
    ↳min_length=5, verbose=False)
trimmed_image = flow.plot_solid_lines(trimmed_lines)
plt.show()

# Add intensity data to lines
line_data = flow.prepare_line_data(trimmed_lines, prepped_intensity, step_size)
angle_data = line_data[2, :, :]
intensity_data = np.array(line_data[4, :, :])
n_dims, max_length, n_lines = line_data.shape
```

Seeding Lines...

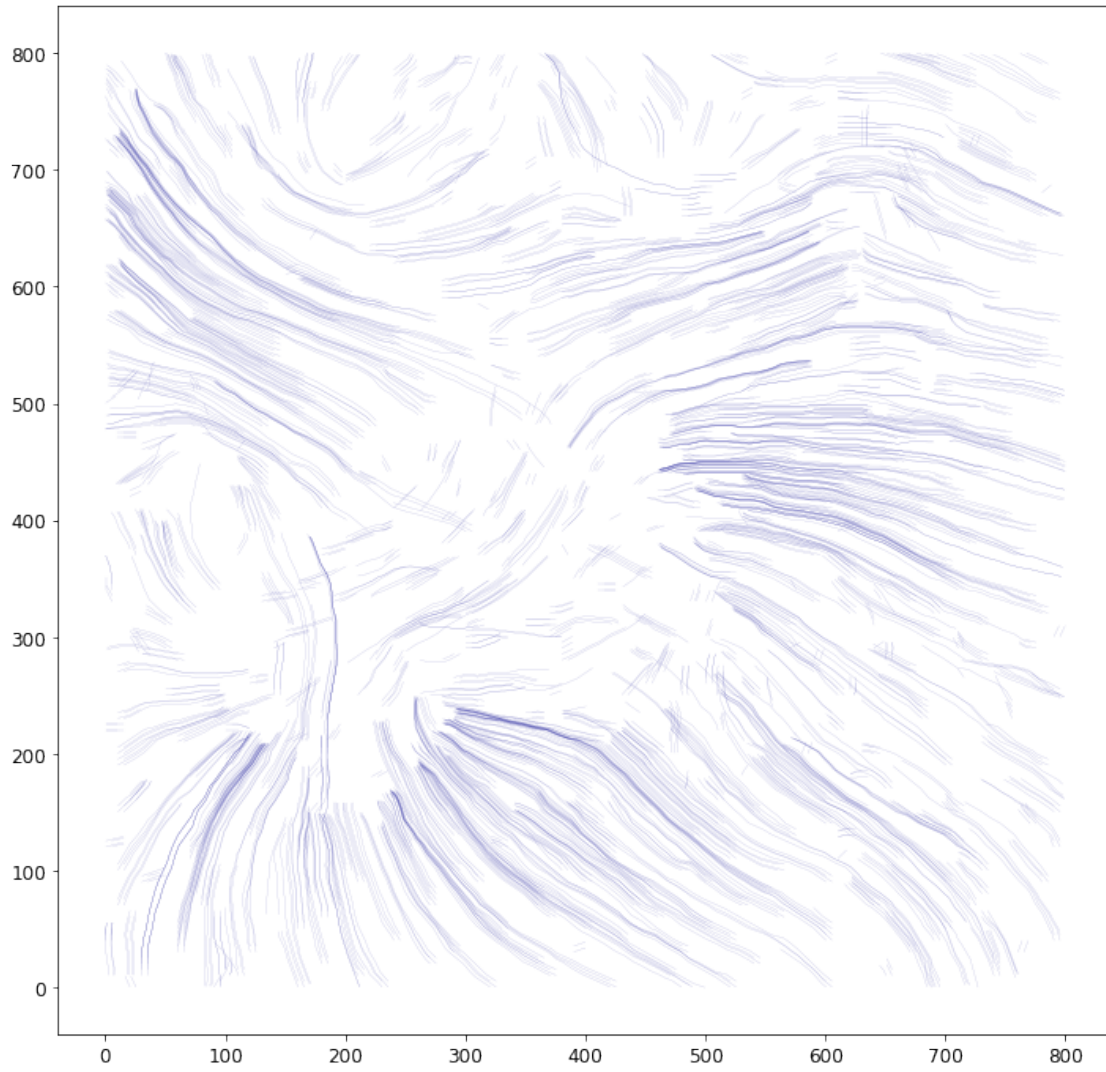
Propogating Lines...

0%

5 %	0:02	remaining
10 %	0:02	remaining
15 %	0:01	remaining
20 %	0:01	remaining
25 %	0:01	remaining
30 %	0:01	remaining
35 %	0:01	remaining
40 %	0:00	remaining
45 %	0:00	remaining
50 %	0:00	remaining
55 %	0:00	remaining
60 %	0:00	remaining
65 %	0:00	remaining
70 %	0:00	remaining
75 %	0:00	remaining
80 %	0:00	remaining
85 %	0:00	remaining
90 %	0:00	remaining
95 %	0:00	remaining

Finished in 0:01

Plotting Solid Lines...

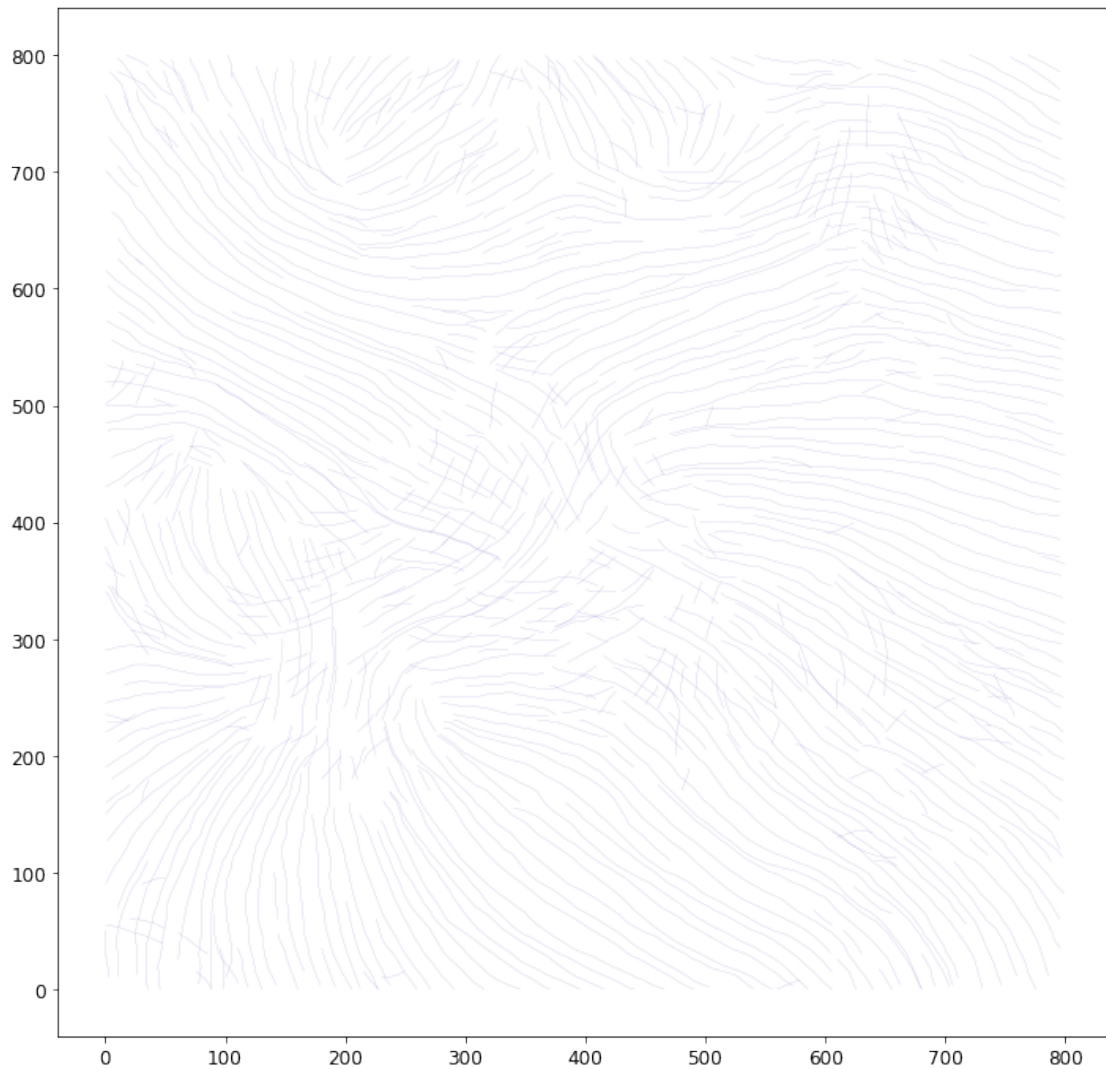


Trimming Lines...

0%

5 %	0:05	remaining
10 %	0:04	remaining
15 %	0:03	remaining
20 %	0:02	remaining
25 %	0:02	remaining
30 %	0:02	remaining
35 %	0:01	remaining
40 %	0:01	remaining
45 %	0:01	remaining
50 %	0:01	remaining
55 %	0:01	remaining
60 %	0:01	remaining

65 % 0:00 remaining
70 % 0:00 remaining
75 % 0:00 remaining
80 % 0:00 remaining
85 % 0:00 remaining
90 % 0:00 remaining
95 % 0:00 remaining
Finished in 0:02
100 % 0:00 remaining
Total Time 0:02
Plotting Solid Lines...



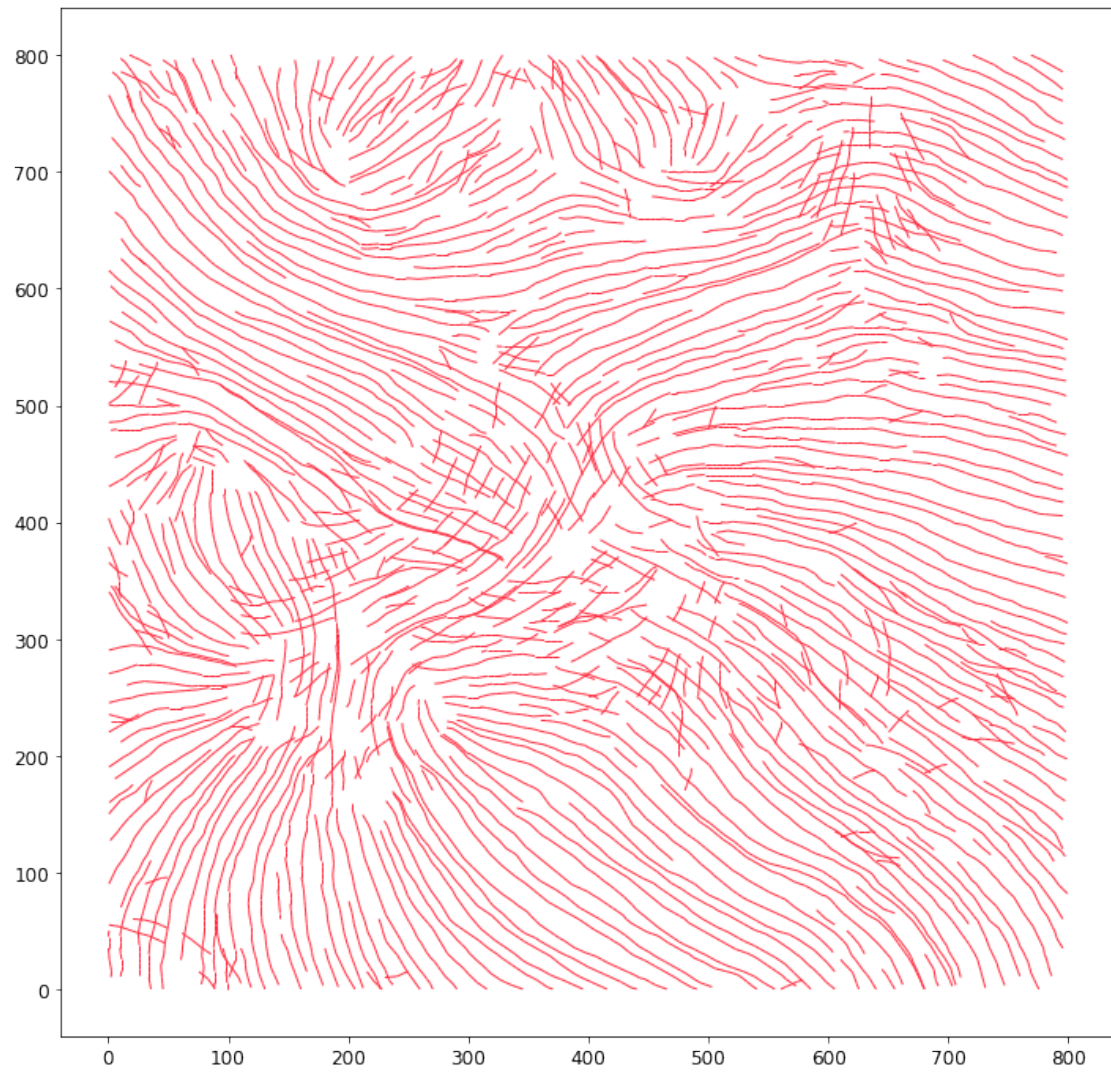
4 Format and Save

```
[7]: # There are many ways to format the plots. I suggest keeping settings
     ↪ organized in the format below.

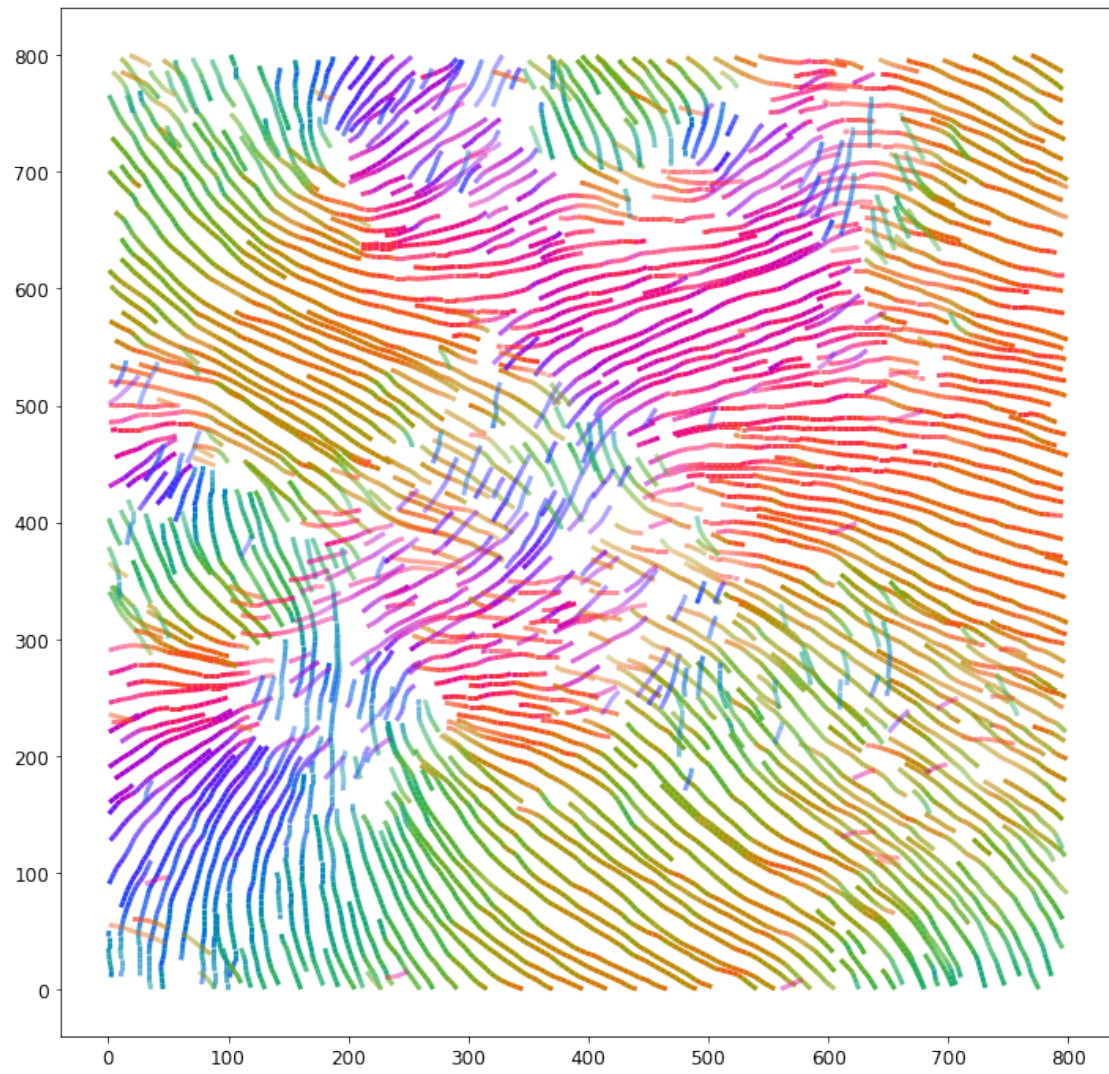
format_codes = [0, 1, 2, 3]
formatted_plots = []
contrast = 3
brightness = 1
gamma = 1
for i, format_code in enumerate(format_codes):
    if format_code == 0:
        # Constant color, linewidth, and alpha
        r, g, b = flow.color_by_angle(np.zeros((max_length, n_lines)))
        linewidth = np.ones((max_length, n_lines)) * 1
        alpha = np.ones((max_length, n_lines))
    elif format_code == 1:
        # Color by angle, alpha by intensity
        alpha = flow.scale_values(flow.smooth(intensity_data, 9),
        ↪ contrast=contrast, gamma=gamma, brightness=brightness)
        linewidth = np.ones((max_length, n_lines)) * 3
        r, g, b = flow.color_by_angle(angle_data)
    elif format_code == 2:
        # Solid Color, alpha by intensity
        r, g, b = flow.color_by_angle(np.zeros((max_length, n_lines)))
        alpha = flow.scale_values(flow.smooth(intensity_data, 9),
        ↪ contrast=contrast, gamma=gamma, brightness=brightness)
        linewidth = np.ones((max_length, n_lines)) * 2
    elif format_code == 3:
        # Faint lines. Good for overlays.
        # Solid Color, alpha by intensity
        shape = np.ones((max_length, n_lines))
        r = shape * 0.95
        g = shape * 0.95
        b = shape * 0.95
        alpha = flow.scale_values(flow.smooth(intensity_data, 9),
        ↪ contrast=contrast, gamma=gamma, brightness=brightness)
        linewidth = np.ones((max_length, n_lines)) * 2

    flow.plot_graded_lines(trimmed_lines, r, g, b, alpha, linewidth)
    formatted_plots.append(plt.gcf())
plt.show()
```

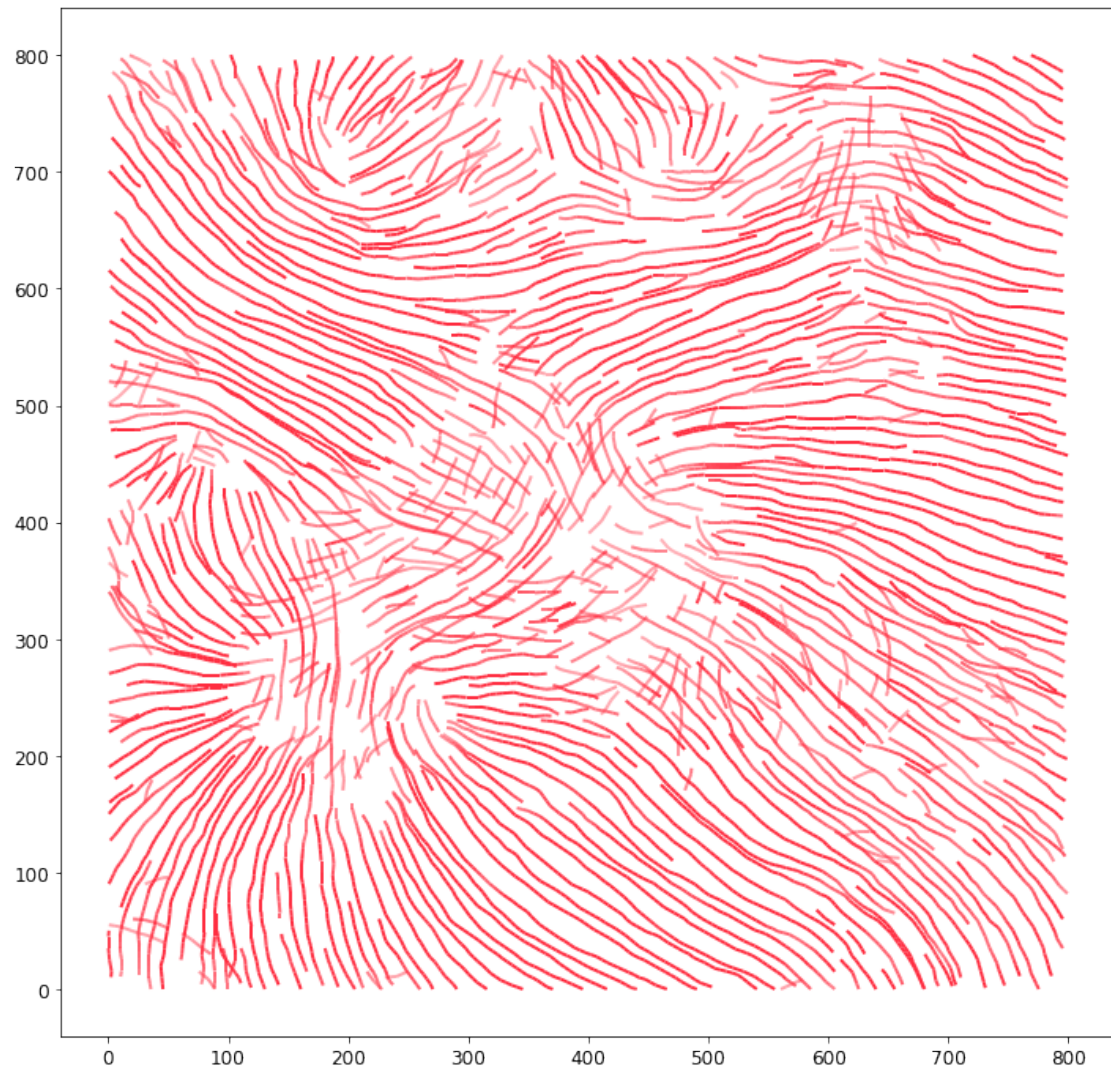
Plotting Graded Lines...



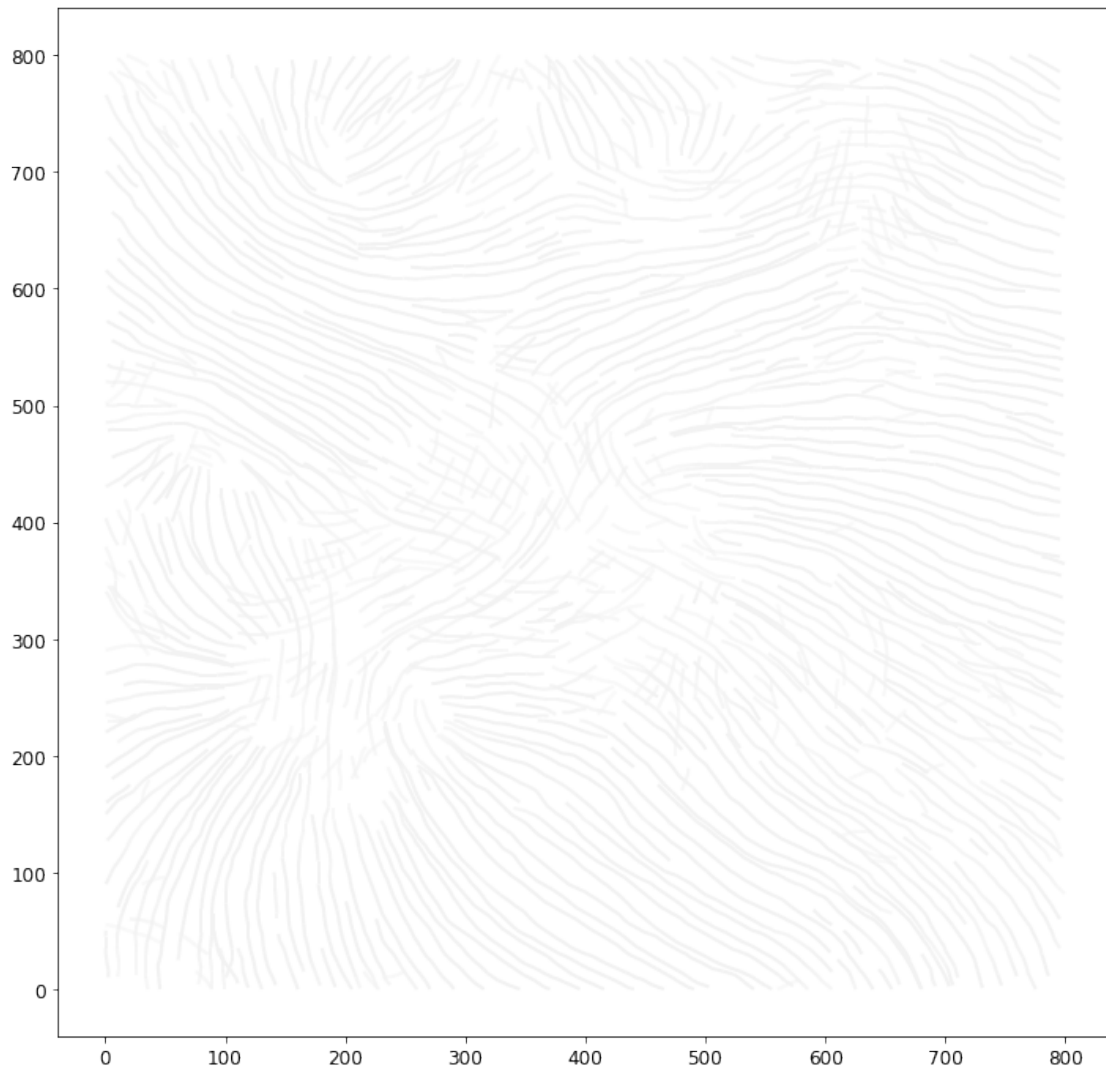
Plotting Graded Lines...



Plotting Graded Lines...



Plotting Graded Lines...



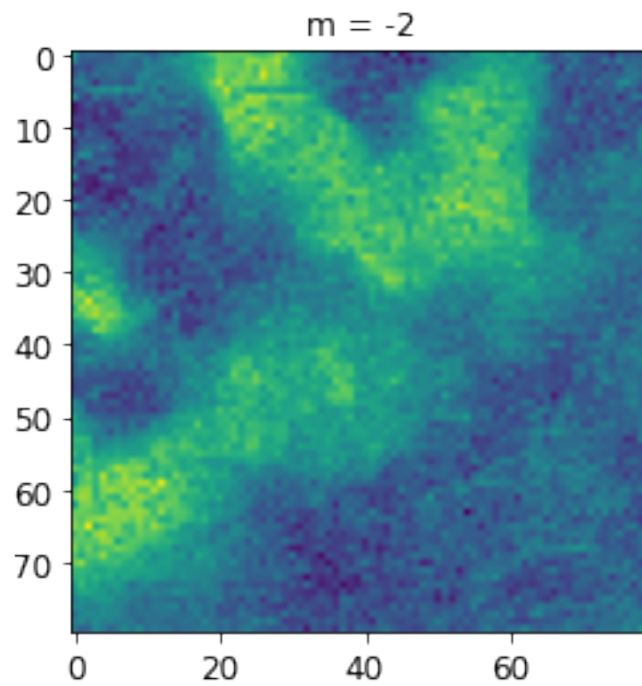
```
[8]: base_filename = "../figures/sample_3_"

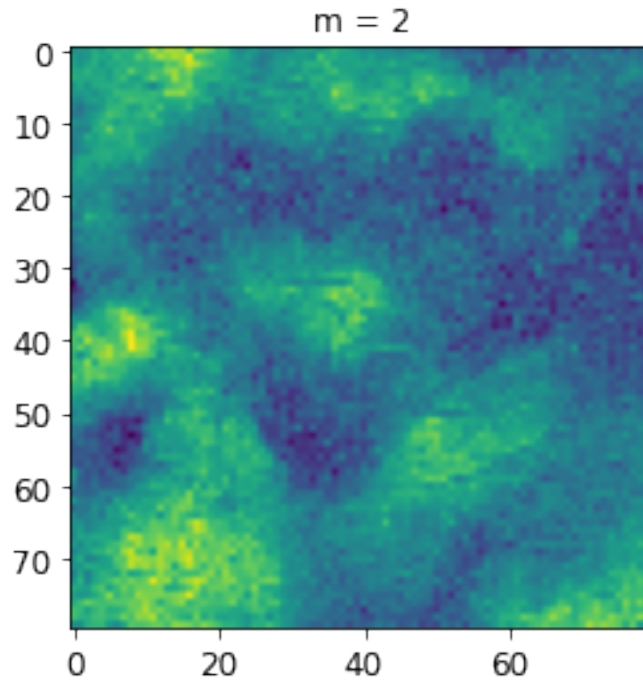
# Save Images
images_to_save = [
    ('propagated_image', propagated_image),
    ('trimmed_image', trimmed_image),
    ('flow_solid', formatted_plots[1]),
    ('flow_color', formatted_plots[2])
]
for name, image in images_to_save:
    image.savefig(base_filename + name + '.png')

formatted_plots[3].savefig(base_filename + 'flow_overlay' + '.png',
    ↪transparent=True)
```


5 Spherical Harmonics

```
[9]: imp.reload(sh)
      harmonics = sh.map_to_spherical_harmonics(integrated_intensity, verbose=True)
```



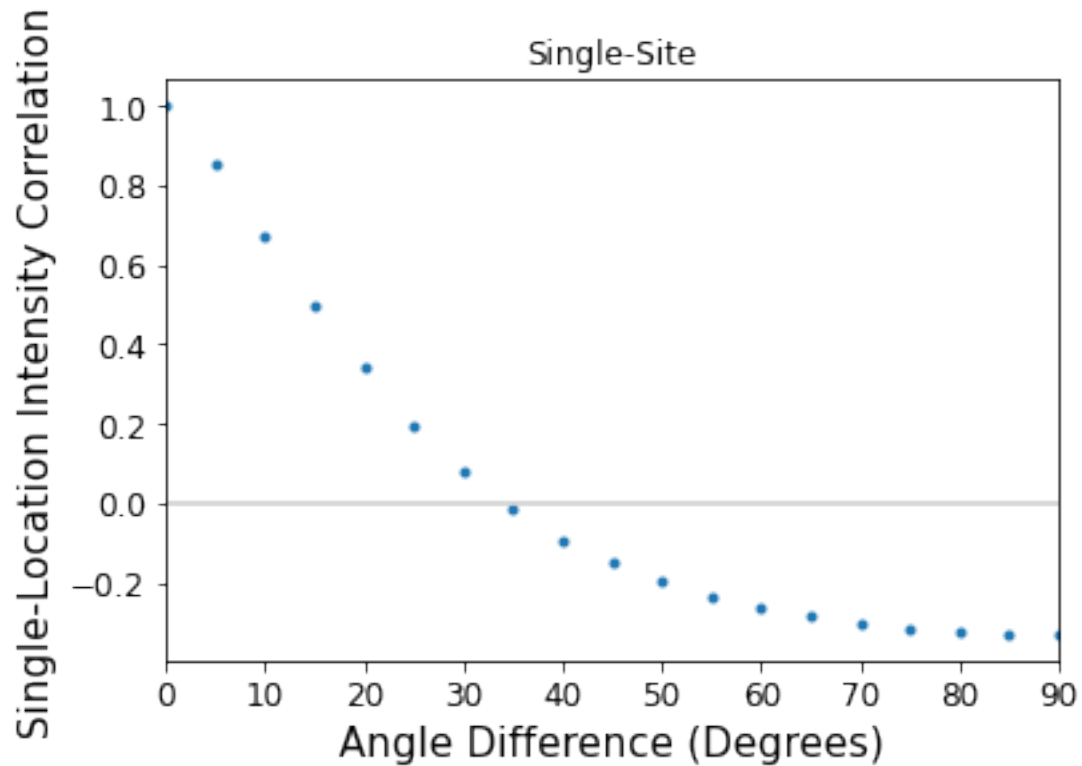


```
[10]: filename = '../data/spherical_harmonics/sample_3_sh_demo.txt'  
      np.savetxt(filename, harmonics, fmt='%.5f')
```

6 Autocorrelation

6.1 Single-Location

```
[11]: imp.reload(ac)  
      figure_image = ac.single_site_autocorrelation(integrated_intensity)
```



6.2 Resultant Autocorrelation

```
[12]: imp.reload(ac)
distances = np.linspace(10, 800, 50, endpoint=True)
n_images, n_angles = integrated_intensity.shape
n_grid = int(np.round(n_images**0.5))
intensity_matrix_resaped = integrated_intensity.reshape(n_grid**2, n_angles)

lc_covariance, stdev, skew, percentiles, scalar_order_parameters = ac.
    ↳resultant_autocorrelation(
        intensity_matrix_resaped, step_size, distances, verbose=False)
```

Setting Up...

Computing Distances...

Computing Angle Differences...

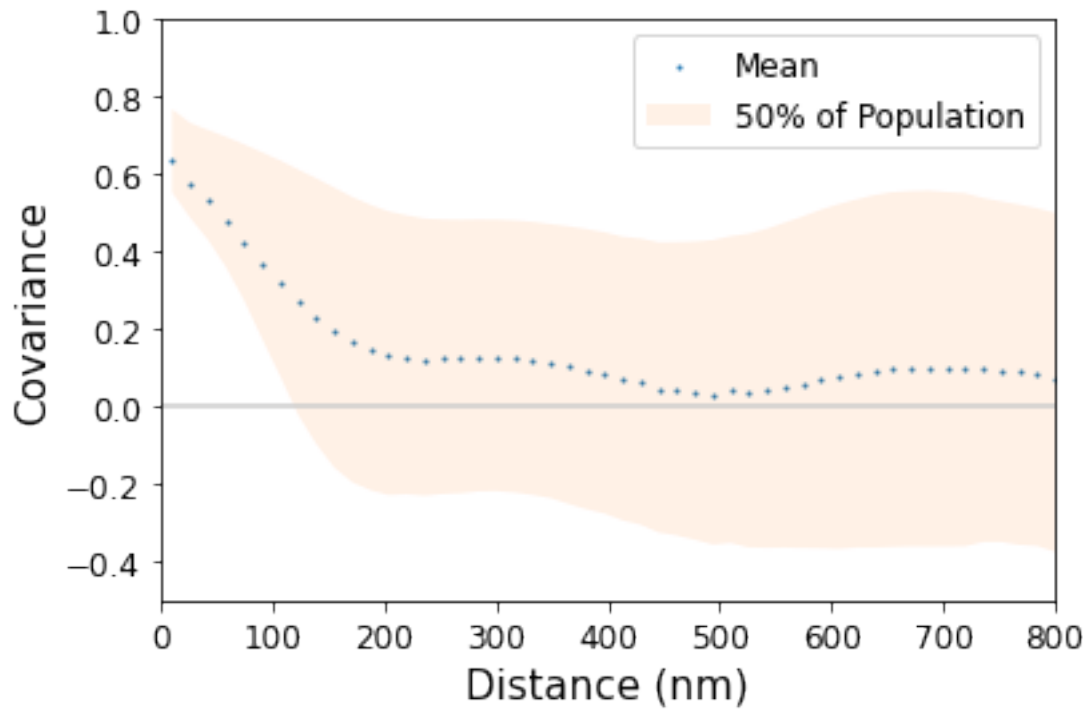
Computing Covariances...

Binning Results...

Dichoric Ratio: 2.3450977795048917

```
[13]: imp.reload(ac)
```

```
lc_covariance_figure = ac.plot_lc_covariance(lc_covariance, percentiles,
↪distances)
plt.show()
```



6.3 Autocorrelation with Angular Distributions

```
[14]: max_grid_distance = 10
grid_distance_step = 1
autocorrelation = ac.distance_angle_autocorrelation(integrated_intensity,
↪max_grid_distance, grid_distance_step, n_slices=10)
```

Setting Up...

0%

Computing data slice number 0...

9 % 0:14 remaining

Computing data slice number 1...

18 % 0:12 remaining

Computing data slice number 2...

27 % 0:11 remaining

Computing data slice number 3...

36 % 0:09 remaining

Computing data slice number 4...


```

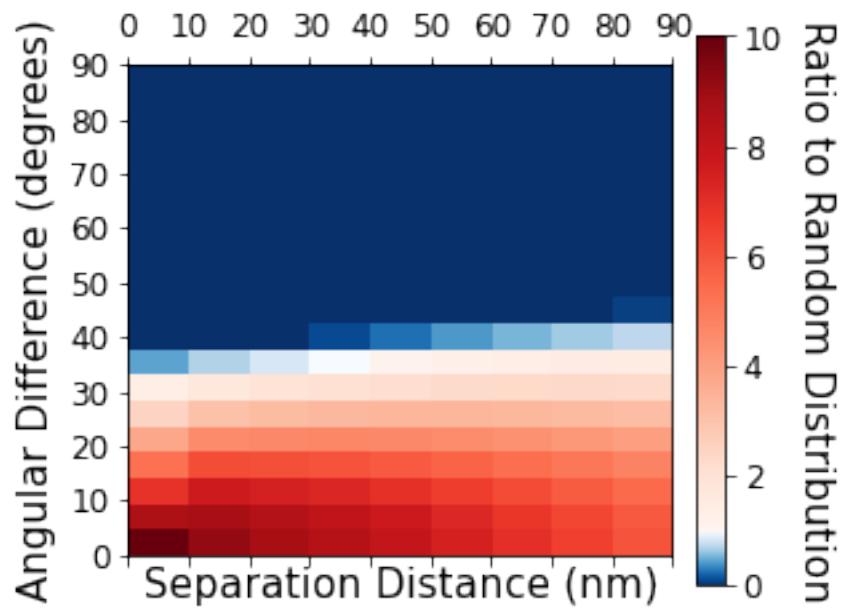
45 %          0:08 remaining
Computing data slice number 5...
54 %          0:07 remaining
Computing data slice number 6...
63 %          0:05 remaining
Computing data slice number 7...
72 %          0:04 remaining
Computing data slice number 8...
81 %          0:02 remaining
Computing data slice number 9...
90 %          0:01 remaining
Finished in 0:13
Wrapping up...
(9, 19)
done

```

```

[15]: image = ac.plot_distance_angle_autocorrelation(autocorrelation, step_size,
↳max_grid_distance, grid_distance_step, interpolate=False)

```



```

[ ]:

```