Import the required libraries In [1]: %matplotlib widget import hyperspy.api as hs import numpy as np import matplotlib.pyplot as plt import atomap.api as am import tkinter as tk from tkinter.filedialog import askdirectory hs.preferences.GUIs.warn_if_guis_are_missing = False hs.preferences.save() plt.rcParams['figure.figsize'] = (8,8) Normalization of the intensity profile Loading the atom lattice: In [2]: root = tk.Tk() root.attributes('-topmost', True) root.iconify() file_path = askdirectory(parent=root) root.destroy() atom_lattice = am.load_atom_lattice_from_hdf5(file_path+'\\data.hdf5',construct_zone_axes=False) sublattice_A=atom_lattice.sublattice_list[0] sublattice_B=atom_lattice.sublattice_list[1] atom_lattice.units=atom_lattice.sublattice_list[0].units atom_lattice.pixel_size=atom_lattice.sublattice_list[0].pixel_size text; selected intensity map, for example: • im_A_high_pass_rl_imag for A intensity map of the RL deconvolution with high-pass image. • im_B_band_pass_pca_imag for B intensity map of the PCA with band-pass image. In [3]: text='im_A_band_pass_pca_imag' intensity_map=np.load(file_path+'\\'+text+'.npy') avg_intensity=np.nanmean(intensity_map[:,:,0],axis=1) std_dev_intensity=np.nanstd(intensity_map[:,:,0],axis=1) avg_axis=np.nanmean(intensity_map[:,:,2],axis=1)*atom_lattice.pixel_size nominal_composition=1 • Intensity profile of the selected intensity map: In [4]: plt.figure(figsize=(12,6)) plt.plot(avg_axis,avg_intensity,'*--') plt.xlabel('Position ['+atom_lattice.units+']') plt.ylabel('Average composition') plt.show() Figure 0.0460 0.0455 Average composition 624000 0.0440 0.0435 10 15 20 25 Position [nm] Histogram to obtain: Mean value of the barriers Mean value of the quantum well In [5]: from scipy.stats import binned_statistic from scipy.signal import find_peaks count_binned=binned_statistic(avg_intensity, avg_intensity, 'count', bins=10) bin_centers=(count_binned[1][1:] + count_binned[1][:-1])/2 mean_binned=binned_statistic(avg_intensity, avg_intensity, 'mean', bins=10) pos_peaks, _ = find_peaks(count_binned[0], height=0) pos_peaks_sorted=pos_peaks[np.argsort(count_binned[0][pos_peaks])] peaks_sorted=mean_binned[0][pos_peaks_sorted] fig, (ax1, ax2) = plt.subplots(2, 1)ax1.plot(avg_axis,avg_intensity,'*--') ax1.set(xlabel='Position ['+atom_lattice.units+']',ylabel='Average intensity') for i in count_binned[1]: ax1.plot(avg_axis,[i]*len(avg_intensity),linestyle='--',color='red',alpha=0.5) _, _, patches = ax2.hist(avg_intensity, bins=count_binned[1], color='lightblue') ax2.plot(bin_centers, count_binned[0], '*--') ax2.set(xlabel='Average intensity',ylabel='Histogram (Count)') fig.tight_layout() plt.show() 0.0460 0.0455 0.0450 Average int 25000 54000 0.0440 0.0435 15 20 25 Position [nm] 35 30 10 0.0455 0.0435 0.0440 0.0445 0.0450 0.0460 Average intensity n_lower_limit: bin index for the lower limit n_upper_limit: bin index for the upper limit In [6]: n_lower_limit=3 n_upper_limit=-2 lower_limit, upper_limit=count_binned[1][n_lower_limit], count_binned[1][n_upper_limit] positions_l=np.where(avg_intensity<lower_limit)</pre> i_barriers=np.nanmean(avg_intensity[positions_l]) positions_u=np.where(avg_intensity>upper_limit) i_quantum_well=np.nanmean(avg_intensity[positions_u]) print('Mean intensity of the barriers: '+str(i_barriers)) print('Mean intensity of the quantum well: '+str(i_quantum_well)) Mean intensity of the barriers: 0.044040000089823034 Mean intensity of the quantum well: 0.04587641168487693 • Selection of the quantum well region with the interfaces after normalization: In [7]: import matplotlib.pyplot as plt from matplotlib.widgets import SpanSelector from matplotlib import gridspec normalized_array=(intensity_map-i_barriers)/(i_quantum_well-i_barriers) avg_norm=np.nanmean(normalized_array[:,:,0],axis=1) std_dev_norm=np.nanstd(normalized_array[:,:,0],axis=1) fig = plt.figure(figsize=(14, 8)) gs = gridspec.GridSpec(1, 2, width_ratios=[2, 1]) ax0 = plt.subplot(gs[0])img1=ax0.plot(avg_norm, '*--') ax0.set(xlabel='Layer',ylabel='Average Composition') def onselect(xmin, xmax): global x_pos x_pos = np.array([xmin,xmax]) span = SpanSelector(ax0, onselect, "horizontal", useblit**=True**, props=dict(alpha=0.5, facecolor="tab:blue"), interactive**=True**, drag_from_anywhere**=True** ax1 = plt.subplot(gs[1])img2=ax1.scatter(intensity_map[:,:,1],intensity_map[:,:,2],s=20,c=normalized_array[:,:,0],cmap='jet',vmin=-0.25,vmax=1.25) fig.colorbar(img2, shrink=0.4, pad=0) ax1.axis('scaled') ax1.axis('off') ax1.set_ylim(ax1.get_ylim()[::-1]) plt.tight_layout() plt.show() Figure 1.0 0.8 - 1.2 0.6 1.0 Average Composition 0.8 - 0.6 0.4 - 0.2 0.2 - 0.0 -0.20.0 20 80 In [8]: muraki_positions=np.arange(x_pos[0]+1,x_pos[1]+1,dtype=int) muraki_signal=avg_norm[muraki_positions] std_dev=std_dev_norm[muraki_positions] sc=hs.signals.Signal1D(muraki_signal) print('Lower layer of the selection: '+str(muraki_positions[0])) print('Upper layer of the selection: '+str(muraki_positions[-1])) Lower layer of the selection: 39 Upper layer of the selection: 50 Fitting (one segregation coefficient): $x(n) = egin{cases} x_0(1-S^n): & 1 \leq n \leq N \ x_0(1-S^N)S^{n-N}: & n > N \end{cases}$ • Build of the Muraki model for one segregation coefficient: In [9]: from hyperspy.component import Component class Muraki(Component): def __init__(self, parameter_1=1, parameter_2=2, parameter_3=3): Component.__init__(self, ('x0', 's', 'N')) self.x0.value = 1self.s.value = 0.5self.N.value = 5self.x0.bmin = 0self.x0.bmax = 1self.s.bmin = 0self.s.bmax = 1self.N.bmin = 0self.N.bmax = 50def function(self, x): x0 = self.x0.values = self.s.value N = self.N.valuereturn np.piecewise(x,[((x >= 1.0) & (x<= N)),x >= N],[lambda x : x0*(1.0 - s**x), lambda x: x0*(1 - s**x)*s**(x-N)]) In [10]: muraki_model = sc.create_model() muraki = Muraki() muraki_model.append(muraki) muraki_model.fit() muraki_model.print_current_values() Model1D: current_component_values: Muraki Active: True Std Parameter Name Free **V**alue Min Max Linear 0.127637 True 1.13936 False 0.638208 0.0579794 0 True False 5.71086 0.430362 False Plot of the fitted model and the input data: In [11]: **from** sklearn.metrics **import** r2_score **def** f(x,x0,s,N): return np.piecewise(x,[((x >= 1.0) & (x<= N)),x >= N],[lambda x : x0*(1.0 - s**x), lambda x: x0*(1 - s**x)*s**(x-N)]) x=np.arange(0,len(sc.data),dtype=float) y_pred=f(x,muraki.x0.value,muraki.s.value,muraki.N.value) r2_parameter=r2_score(sc.data[0::], y_pred[0::]) from matplotlib.offsetbox import AnchoredText plt.figure() plt.plot(avg_axis[np.arange(0,len(avg_intensity))],avg_norm,'*--') plt.plot(avg_axis[x.astype(int)+muraki_positions[0]],y_pred,'-',color='red') plt.xlabel('Position [nm]') plt.ylabel('Average Composition') plt.minorticks_on() plot=plt.gca() $label='$R^2 = $'+str(np.round(r2_parameter,3))$ at = AnchoredText(label, prop=dict(size=10), frameon=True, loc='upper right') at.patch.set(edgecolor='lightgray') at.patch.set_boxstyle('round, pad=0., rounding_size=0.1') plot.add_artist(at) plt.show() Figure $R^2 = 0.905$ 1.0 0.6 Average Composition 20 25 15 Position [nm] In [12]: plt.savefig(file_path+'\\muraki_1s_'+text+'.png',dpi=300,transparent=True,bbox_inches='tight') Fitting (two segregation coefficients): $x(n) = \left\{ egin{array}{ll} x_0(1-S_l^n) : & 1 \leq n \leq N \ x_0(1-S_l^N)S_u^{n-N} : & n > N \end{array}
ight.$ Build of the Muraki model for two segregation coefficients: In [13]: class Muraki2(Component): def __init__(self, parameter_1=1, parameter_2=2, parameter_3=3, parameter_4=4): Component.__init__(self, ('x0', 's1', 's2', 'N')) self.x0.value = 1self.s1.value = 0.5self.s2.value = 0.5self.N.value = 5 self.x0.bmin = 0self.x0.bmax = 1self.s1.bmin = 0self.s1.bmax = 1self.s2.bmin = 0self.s2.bmax = 1self.N.bmin = 0self.N.bmax = 50def function(self, x): x0 = self.x0.values1 = self.s1.values2 = self.s2.valueN = self.N.valuereturn np.piecewise(x,[((x >= 1.0) & (x<= N)),x >= N],[lambda x : x0*(1.0 - s1**x), lambda x: x0*(1 - s2**x)*s2**(x-N)])In [14]: muraki_model = sc.create_model() muraki = Muraki2() muraki_model.append(muraki) muraki_model.fit() muraki_model.print_current_values() Model1D: current_component_values: Muraki2 Active: True Value Std Min Free Max Parameter Name Linear True 1.49528 0.61454 False s1 0.769065 0.139011 0 True False 0.613038 0.0653428 False 50 5.20866 0.902382 0 False Plot of the fitted model and the input data: In [15]: def f(x, x0, s1, s2, N): return np.piecewise(x,[((x >= 1.0) & (x<= N)),x >= N],[lambda x : x0*(1.0 - s1**x), lambda x: x0*(1 - s2**x)*s2**(x-N)])x=np.arange(0,len(sc.data),dtype=float) y_pred=f(x,muraki.x0.value,muraki.s1.value,muraki.s2.value,muraki.N.value) r2_parameter=r2_score(sc.data[1::], y_pred[1::]) plt.figure() plt.plot(avg_axis[np.arange(0,len(avg_intensity))],avg_norm,'*--') plt.plot(avg_axis[x.astype(int)+muraki_positions[0]],y_pred,'-',color='red') plt.xlabel('Position [nm]') plt.ylabel('Average Composition') plt.minorticks_on() plot=plt.gca() $label='$R^2 = $'+str(np.round(r2_parameter,3))$ at = AnchoredText(label, prop=dict(size=10), frameon=True, loc='upper right') at.patch.set(edgecolor='lightgray') at.patch.set_boxstyle('round, pad=0., rounding_size=0.1') plot.add_artist(at) plt.show() Figure $R^2 = 0.925$ 1.0 0.8 Average Composition 0.2 10 15 20 25 Position [nm] In [16]: plt.savefig(file_path+'\\muraki_2s_'+text+'.png',dpi=300,transparent=True,bbox_inches='tight')