DSpectra_scripts

February 8, 2018

```
In [1]: # Import modues
        # -*- encoding: utf-8 -*-
        import numpy as np
        import matplotlib as mpl
        from matplotlib import rc
        import math
        import pandas as pd
        import os
        import itertools
        from scipy import stats
        from scipy import ndimage
        import seaborn as sns
        import matplotlib as mpl
        from matplotlib import cm
        import matplotlib.pyplot as plt
        from mpl_toolkits.mplot3d import Axes3D
        from statsmodels.stats.descriptivestats import sign_test
        from statsmodels.stats.weightstats import zconfint
        from statsmodels.stats.weightstats import *
        from skimage import measure
        from scipy import ndimage
        from scipy import misc
        from scipy.stats.stats import pearsonr, spearmanr
        from collections import Counter
        # from pandas import ExcelWriter
        from sklearn.cross_validation import train_test_split
        from sklearn.linear_model import LinearRegression
        from sklearn.linear_model import Lasso
        from sklearn.linear_model import Ridge
        from sklearn.metrics import mean_squared_error
        from sklearn.ensemble import RandomForestRegressor
        import statsmodels.stats.api as sm
```

```
from sklearn.naive_bayes import MultinomialNB
        from sklearn.ensemble import RandomForestClassifier
        from sklearn.preprocessing import OneHotEncoder
        from sklearn.cluster import KMeans
        from sklearn.neighbors import KNeighborsClassifier
        from sklearn.metrics import classification_report
        from sklearn import cross_validation, datasets, grid_search, linear_model, metrics
        from sklearn.metrics import classification_report
        from scipy.optimize import curve_fit
        from scipy import signal
        from sklearn import random_projection
        from sklearn.decomposition import RandomizedPCA
        from sklearn.decomposition import PCA
        from sklearn import manifold
        from sklearn.cluster import KMeans
        sns.set_style("whitegrid")
        # flatui = ["#9b59b6", "#3498db", "#95a5a6", "#e74c3c", "#34495e", "#2ecc71"]
        sns.set_palette('Accent')
        rc('font', family='Arial', size=14) # change font for russian
        rc('xtick', labelsize=14)
        rc('ytick', labelsize=14)
        mpl.rcParams.update({'font.size': 14})
        % matplotlib inline
        print 'Import Ready'
Import Ready
/usr/local/lib/python2.7/dist-packages/sklearn/cross_validation.py:41: DeprecationWarning: This
  "This module will be removed in 0.20.", DeprecationWarning)
/usr/local/lib/python2.7/dist-packages/statsmodels/compat/pandas.py:56: FutureWarning: The panda
  from pandas.core import datetools
/usr/local/lib/python2.7/dist-packages/sklearn/grid_search.py:42: DeprecationWarning: This modul
  DeprecationWarning)
```

from sklearn import cross_validation, datasets, linear_model, metrics

1 Spectrum parameters

```
cmap = plt.get_cmap('Accent')
        colors = [cmap(i) for i in np.linspace(0., 1., 8)] * 2
In [3]: # Function for spectra features
        def smooth (spec_i, params, type_smooth):
            # type 1 - average filter, type 2 - SavGol, type 3 - FFT
            Function for smoothing spectra. Return smooth spectra as np.array of intencity (np.a
            Parameters:
                spectra (list of intence)
                parameters of smooth (optimizing):
                    type 1: medians - [window (full size, int)]
                    type 2: average - [window (full size, int)]
                    type 3: SavGol - [polinom order (int), windows (full size, odd, int)]
                    type 4: Discret Furie - [min frequence (float [0, 2])]
                type smooth (see above)
            , , ,
            spec_i = np.array(spec_i)
            spec_tranc = []
            if type_smooth == 0:
                # median filter w - full window size
                window = params[0]
                spec_i = np.array(spec_i)
                for i in xrange(len(spec_i)):
                    d = None
                    if i < window:</pre>
                        d = np.median(spec_i[0:i+window])
                    elif i + window > len(spec_i):
                        d = np.median(spec_i[i-window:])
                    else:
                        d = np.median(spec_i[int(i-window/2):int(i+window/2)])
                    spec_tranc.append(d)
            elif type_smooth == 1:
                # average filter w - full window size
                window = params[0]
                spec_i = np.array(spec_i)
                for i in xrange(len(spec_i)):
                    d = None
                    if i < window:</pre>
                        d = np.average(spec_i[0:i+window])
                    elif i + window > len(spec_i):
                        d = np.average(spec_i[i-window:])
                    else:
                        d = np.average(spec_i[int(i-window/2):int(i+window/2)])
                    spec_tranc.append(d)
            elif type_smooth == 2:
                # Sav-Gol filter
```

```
rate = 1
        order = params[0]
        window_size = params[1] # full size
        if window_size % 2 != 1 or window_size < 1:</pre>
            print 'window_size size must be a positive odd number'
        if window_size < order + 2:</pre>
            print 'window_size is too small for the polynomials order'
        order_range = xrange(order+1)
        half_window = (window_size - 1) // 2
        b = np.mat([[k**i for i in order_range] for k in xrange(- half_window, half_window)
        m = np.linalg.pinv(b).A[deriv] * rate**deriv * math.factorial(deriv)
        firstvals = spec_i[0] - np.abs(spec_i[1:half_window+1][::-1] - spec_i[0])
        lastvals = spec_i[-1] + np.abs(spec_i[-half_window-1:-1][::-1] - spec_i[-1])
        spec_i = np.concatenate((firstvals, spec_i, lastvals))
        spec_tranc = np.convolve(m[::-1], spec_i, mode='valid')
        lendelta = len(spec_tranc) - len(spec_i)
        if lendelta > 0:
            print 'lendelta: ', lendelta
            spec_tranc = spec_tranc[lendelta // 2:]
    else:
        # Calculate furie-conversion os spec
        fft_res = np.fft.fft(spec_i)
        u_crit = params[0]
        n = float(len(fft_res)) # number of channels
        filtered = map(lambda x: 0 if (x / n > u_crit) else fft_res[x], xrange(len(fft_r
        spec_tranc = map(lambda x: x.real, np.fft.ifft(filtered))
    return np.array(spec_tranc)
# Autofind peacks and base line
def calc_h(i, w, v):
    Calulate h parameters for 'zero-filter' algorithm. rturn h value (float).
    Parameters:
        i - channel value [0, 4095] (int)
        w - windows (even, int)
        v - windows shift (int)
    111
    h = None
    if (-v - w / 2) \le i \le -w / 2:
        h = -0.5 / v
    elif -0.5*w \le i \le 0.5*w:
```

deriv = 0

```
h = 1. / w
    elif 0.5*w < i <= (0.5*w+v):
        h = -0.5 / v
    else:
        print 'Not good: ', i
    if h == 0: print 'WTF'
    return h
def zero_area(spec, params, type_alg):
    # type_alg = 1 - classical zero-area algorithm, 2, 3 - my algo as float average and
    Function for 'zero-area'. Return mask of base line as np.array of baseline intence (
    Parameters:
        spectra (list of intence)
        parameters (optimizing):
            type 1: classical - [window (even, int), window_shift (int)]
            type 2: my as average - [window (int)] - work bad
            type 3: my as parabola - [windows (int)] - work bad
        type smooth (see above)
    , , ,
    spec_transform = []
    spec = np.array(spec)
    if type_alg == 1:
        w = params[0]
        v = params[1]
        if w % 2 != 0:
            print 'w mast be even' #
            return
        for i in xrange(len(spec)):
            i = int(i)
            d = 0
            p1 = -v-0.5*w
            p2 = +v+0.5*w
            j = int(p1)
            if i + p1 < 0:
                while j != p2:
                    if i + j < 0:
                        d += calc_h(j, w, v) * spec[0]
                         d \leftarrow calc_h(j, w, v) * spec[int(i+j)]
                    j+=1
            elif i + p2 > len(spec):
                while j != p2:
                    if i + j \ge len(spec):
                         d += calc_h(j, w, v) * spec[-1]
                    else:
                         d \leftarrow calc_h(j, w, v) * spec[int(i+j)]
```

```
j+=1
        else:
            while j != p2:
                d \leftarrow calc_h(j, w, v) * spec[int(i+j)]
                j+=1
        spec_transform.append(d)
elif type_alg == 2:
    # Can write the same algo for smoothind filters type 0-2
    chanel = xrange(len(spec))
    w = params[0] # windows full-lenght
    spec_transform = range(len(spec))
    num_w = int(len(spec) / w) # numbers of subspec for find peaks
    is_peak_up = [0] * num_w
    is_peak_out = [0] * num_w
    spec_parts = []
    # write cond for peak up and down
    for i in xrange(num_w):
        start_i = w * i
        stop_i = w * (i + 1)
        if stop_i > len(spec): stop_i = len(spec)
        spec_i = np.array(spec[start_i: stop_i])
        spec_parts.append(spec_i)
        aver_i = np.average(spec_i) # may be use median? - , ...
        cond_i = spec_i > aver_i
        if np.sum(cond_i) > w / 2: # cond for peak
            if np.sum(cond_i[:w/2]) > np.sum(cond_i[w/2:]):
                is_peak_out[i] = 1
            else:
                is_peak_up[i] = 1
    for i in xrange(num_w-3):
        d = is_peak_up[i]
        if d and np.sum(is_peak_out[i:i+3]):
            for j in xrange(len(spec_parts[i])):
                spec_transform[i+j] = 0
        else:
            for j in xrange(len(spec_parts[i])):
                spec_transform[i+j] = spec_parts[i][j]
    spec_transform.append([0 for x in xrange((num_w-3)*w)])
elif type_alg == 3:
    deriv = 0
    rate = 1
    order = 4
    window_size = params[0] # full size
    if window_size < order + 2:</pre>
        print 'window_size is too small for the polynomials order'
        return
    chanel = xrange(len(spec))
```

```
spec_transform = range(len(spec))
        num_w = int(len(spec) / window_size) # numbers of subspec for find peaks
        spec_parts = [spec[x*window_size:(x+1)*window_size] for x in xrange(num_w-1)]
        for i in xrange(len(spec_parts)):
            # aprox point as 2nd order polinom and calculate a and R2
            raw_spec = spec_parts[i]
            coef, covar, R2 = approx_poly(xrange(len(raw_spec)), raw_spec, 2)
            a = coef[0]
            if a < 0 and R2 > 0.7:
                # is peak
                for j in xrange(len(spec_parts[i])):
                    spec_transform[i+j] = 0
            else:
                for j in xrange(len(spec_parts[i])):
                    spec_transform[i+j] = spec_parts[i][j]
    return np.array(spec_transform[:4095])
def is_baseline(spec_tr, floor = -1):
    # return True/false mask for baseline
    111
    Function for baseline mask. Return mask as np.array of boolen (np.array)
    Parameters:
        spectra of baseline (list of intence)
        parameters (optimizing):
            medians - [flag of maximum base intence (int or float)]
    spec_tr = np.abs(spec_tr)
    flag = floor
    if floor < 0:
        flag = np.average(spec_tr)
    spec_res = spec_tr < flag</pre>
    for i in xrange(10,len(spec_res)-10):
        if spec_res[i] and np.sum(spec_res[i-10:i+10]) < 10:</pre>
            spec_res[i] = False
    return spec_res
# Calculate finish parameters
def approx_poly(list_x, list_y, order):
    # calculate coef for polinom approximation [A, B], y = Ax + B return highest power f
    Function for approximate points as order poinom. Return approximate coeffisients? co
    Parameters:
        list of x points
        list of y points
        order of polinom
```

```
111
    poly_coeff = np.polyfit(list_x, list_y, order, cov=True)
    R2 = None
    # r-squared for 1 order TODO: add Matius coefficient
    if (order - 1.0) < 1.5: # use 0.5 for 1st order polinom
        p = np.poly1d(poly_coeff[0])
        # fit values, and mean
                                                 # or [p(z) for z in x]
        yhat = p(list_x)
        ybar = np.sum(list_y)/len(list_y)
                                                   # or sum(y)/len(y)
        ssreg = np.sum((yhat-ybar)**2) # or sum([ (yihat - ybar)**2 for yihat in yhat]
        sstot = np.sum((list_y - ybar)**2) # or sum([(yi - ybar)**2 for yi in y])
        R2 = ssreg / sstot
        R2 = round(float(R2), 5)
    return poly_coeff[0], poly_coeff[1], R2
def cur_poly_value(x, coef):
    # return y for x and coef of polinom, coef highest power first
    Function for calculate point with polinoms coefficients. Return value of polinom (fl
    Parameters:
       point (int)
        coefficient (highest pover first, list)
   res = 0.0
    cur_x_dg = 1.0
    for i in xrange(1, len(coef)+1):
        res += cur_x_dg * coef[-i]
        cur_x_dg *= x
    return res
def calc_all_area(spec_i):
    # return area of spectra
    Function for calculate area of baseline. Return data (float).
    Parameters:
        baseline spectra of intencity (list)
    return np.sum(spec_i)
def calc_baseline(spec_i, baseline_mask):
    # return Imax ans S of background
    Function for approximate baseline as 5th order poinom. Return approximate baseline a
    Parameters:
        spectra (list of intence)
```

```
baseline mask (list of boolen)
    spec_i = np.array(spec_i)
    x = np.array([x for x in xrange(len(spec_i))])
    poly_coef = approx_poly(x[baseline_mask], spec_i[baseline_mask], 5)[0]
    approx_baseline = []
    for i in x:
        approx_baseline.append(cur_poly_value(i, poly_coef))
    approx_baseline = np.array(approx_baseline)
    return approx_baseline
def calc_peacs(spec_i, base_fit, baseline_mask, parameters = None, type_calc=2, energy =
    # return i_max, S and channel of peack
    Function for calculate peak data.
    Return:
        intencity of peaks (list)
        cannels of peaks (list)
        gauss intencity of each peaks (list)
        gayss squares of each peaks (list)
        intencity of fitting base line (list)
    Parameters:
        spectra (list of intence)
        baseline (list of intence)
        baseline mask (list of boolen)
        parameters (for optimisation):
            type 1 (derivative): [iteration for smooth of diff spectra [0, 10], treshold
            type 2 (zero-area base): [iteration of smooth (int)]
        type (see above)
        energy of close peak (treshold parameters, float 0.2)
        a0, a1 - channel to energy parameters (float)
    ,,,
    spec_i = np.array(spec_i)
    max_delta_chan = None
    if a0 is not None and a1 is not None:
        max_delta_chan = (energy - a0) / a1 # for close peaks
    else:
       max_delta_chan = (energy - spe_df['a0'][0]) / spe_df['a1'][0] # for close peaks
    if type_calc == 1:
        # derivative type - for all spectra
        itera = parameters[0] #
        der_threshold = parameters[1] # 10 for work - oprimize
        spec_new = np.diff(spec_i, 2)
        while itera > 0:
            spec_new = smooth(spec_new, params=[3, 11], type_smooth=2) # Sav-Gol
            itera -= 1
```

```
der = np.array(spec_new)
    peak_mask = der < der_threshold # peak search</pre>
    i_all = []
    channel_all = []
    delta_peak = 0
    for i in xrange(len(peak_mask)-1):
        # search peaks channel
        if peak_mask[i] and peak_mask[i+1]: # TODO: set threshold longer then i+1?
            # if peak prolong
            delta_peak += 1
        elif peak_mask[i]:
            # end peak
            if delta_peak > 0:
                i_max = np.max(spec_i[i-delta_peak:i]) # max intence
                c = np.where(spec_i[i-delta_peak:i] == i_max)[0][0] # channel of mo
                i_max = np.average(spec_i[i-delta_peak+c-3:i-delta_peak+c+3]) # call
                i_all.append(i_max)
            else:
                pass
        else:
            # no peak
            delta = 0
    # calculate square of peak and (I - base line)
    i_all = np.array(i_all)
    gauss_i, gauss_s, base_i = gauss_peaks(i_all, channel_all, base_fit, spec_i)
elif type_calc == 2:
    w0 = 3 # parameters[0] # SG polinom order
    w1 = 11 # parameters[1] # SG window size
    itera = parameters[0] # numbers of iteration - optimize, 2
    x = np.array(xrange(len(spec_i)))
    x_peak = x[~baseline_mask] # mask for peaks
    spec_new = []
    for i in xrange(len(spec_i)):
        # for new 'spec' with peak only
        if i in x_peak:
            spec_new.append(spec_i[i])
        else:
            spec_new.append(0)
    while itera > 0:
        spec_new = smooth(spec_new, params=[w0, w1], type_smooth=2) # Sav-Gol
        itera -= 1
    channel_all = (np.diff(np.sign(np.diff(spec_new))) < 0).nonzero()[0] # + 1 # la</pre>
    spec_new = np.array(spec_new)
    i_all = spec_new[channel_all]
    i_all = np.array(i_all)
    channel_all = channel_all[i_all > 1.]
```

```
i_all = i_all[i_all > 1.]
        gauss_i, gauss_s, base_i = gauss_peaks(i_all, channel_all, base_fit, spec_i)
    return i_all, channel_all, gauss_i, gauss_s, base_i
def gaus(x,a,x0,sigma):
    # for gaussian fit
    Function for gaussian curve. Return gauss data (int or np.array)
    Parameters:
        x point
        aplitude
        senter of peak
        std of peak
    return a * np.exp(-(x - x0)*(x - x0) / (2 * sigma*sigma))
def gauss_peaks(i_all, channel_all, base_fit, spec_i):
    # calculate Gauss S and I of peaks
    Function for gaussian approximation of peaks.
    Return:
        intens of peaks (np.array)
        squares of peaks (np.array)
        intence of base line (np.array)
    Parameters:
        intence of peaks (list)
        channels of peaks (list)
        intence of fit baseline (list)
        intence of raw spectra (list)
    ,,,
    base_i = []
    gauss_s = []
    gauss_i = []
    for i in xrange(len(channel_all)):
        c_middle = channel_all[i]
        base_i.append(np.average(base_fit[c_middle]))
        g_h = 0
        g_s = 0
        w = 15
        y = np.array(spec_i[c_middle-w:c_middle+w]) # gauss fit of points
        x = np.array(xrange(len(y)))
        if y.shape is not () and x.shape is not ():
        # correction for weighted arithmetic mean
            mean = np.sum(x * y) / np.sum(y)
            sigma = np.sqrt(np.sum(y * (x - mean)**2) / np.sum(y))
            try:
```

```
popt,pcov = curve_fit(gaus, x, y, p0=[max(y), mean, sigma])
                        y_fit = gaus(x, *popt)
                        g_h = np.max(y_fit)
                        g_s = g_h * (2 * np.pi) ** 0.5 * popt[-1] # gauss square
                        w += 2
                    except RuntimeError as var:
                        g_h = 0
                        g_s = 0
                    except ValueError as var:
                        print var
                        print x, y
                        g_h = 0
                        g_s = 0
                else:
                    g_h = 0
                    g_s = 0
                gauss_i.append(g_h)
                gauss_s.append(g_s)
            return np.array(gauss_i), np.array(gauss_s), np.array(base_i)
In [4]: # load spec data
        def dir_name(_):
            _{-} = _{[0].split('\\')}
            if len(_) > 1: return 'specs/' + _[1]
        dirs = ['specs/' + x for x in list(os.walk('specs'))[0][1]]
        dict_obj_spe = {}
        directs = dirs
        shifr = dict(zip(range(len(directs)), directs))
        counter = 0
        for dir_name in directs:
            if dir_name:
                print 'Files in directory: ', dir_name, len(os.listdir(dir_name))
                data_num = xrange(len(os.listdir(dir_name)))
                n = dir_name.split('/')[1]
                n = n.split('_')
                ftype = n[0].lower()
                mark = n[1]
                fraction =n[2]
                kd = n[3] # binary yes/no
                if len(kd.split('.')) > 1:
                    kd = kd.split('.')[1] # float or none - undefine
```

```
for i in xrange(len(os.listdir(dir_name))):
    dict_obj_spe[counter] = {}
    i = os.listdir(dir_name)[i]
    name = str(i)
    if '.spe' in i:
        spec_file = open(dir_name + '/' + i)
        spec_num = i.split('_')[-1][0] # select one number
        if spec_num in ['1', '2']:
            _ = i.split('_')[-1][1]
            if _ not in ['(', '.']:
                spec_num += _
        if ftype == 'npss.zn':
            spec_num = i.split('_')[-2] + '_' + spec_num
        if ftype == 'an':
            spec_num = '_'.join(i.split('_')[1:4])
            spec_num = spec_num[:-4]
        if spec_num in ['(','.','00','k','n']:
            print i, ' ', spec_num, ' ', ftype
        lines = spec_file.readlines()
        spec_file.close()
        # properties
        exposition = None
        voltage = None
        anod_current = None
        atmo = None
        list_e = []
        list_i = []
        try:
            exposition = int(lines[3])
            voltage = int(lines[6])
            anod_current = int(lines[7])
            atmo = int(lines[17])
        except ValueError as var:
            print ('ERROR: ', var)
        a0 = float(lines[19])
        a1 = float(lines[20])
        a2 = float(lines[21])
        lines_spec = tuple(lines[32:])
        c = 0
        for ind in lines_spec:
            e = a0 + a1*c + a2*c*c
            list_e.append(e)
```

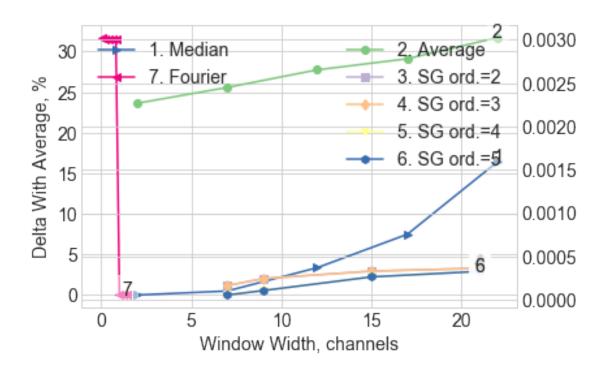
```
list_i.append(int(ind))
                            c += 1
                        dict_obj_spe[counter]['exposition'] = exposition
                        dict_obj_spe[counter]['voltage'] = voltage
                        dict_obj_spe[counter]['current'] = anod_current
                        dict_obj_spe[counter]['atmo'] = atmo
                        dict_obj_spe[counter]['ftype'] = ftype
                        dict_obj_spe[counter]['mark'] = mark
                        dict_obj_spe[counter]['spec_num'] = spec_num
                        dict_obj_spe[counter]['fraction'] = fraction
                        dict_obj_spe[counter]['kd'] = kd
                        dict_obj_spe[counter]['a0'] = a0
                        dict_obj_spe[counter]['a1'] = a1
                        dict_obj_spe[counter]['a2'] = a2
                        dict_obj_spe[counter]['energy'] = list_e
                        dict_obj_spe[counter]['intence'] = list_i
                        counter += 1
        print ('Done')
Files in directory: specs/NPKS_4.30.15.16_rawgrain_kd.None 10
Files in directory: specs/NPKS_4.30.15.16_grain_kd.None 24
Files in directory: specs/NPKS_4.30.15.16_100_kd.None 29
Files in directory: specs/NPKS_4.30.15.16_500_kd.None 26
Files in directory: specs/NPKS_4.30.15.16_500.dry_kd.None 15
Files in directory: specs/NPKS_4.30.15.16_100.dry_kd.None 14
Done
In [5]: # Create df
        spe_df = pd.DataFrame.from_dict(dict_obj_spe, orient='index')
1.1
In [6]: # smoothing spectra
        test_df = spe_df[spe_df.fraction == 'grain']
        ftype_all = Counter(test_df.mark)
        print ftype_all
        aver_spec_type = {}
        smooth_type = {
            0: xrange(2, 23, 5),
            1: xrange(2, 23, 5),
            2: [xrange(2,6), (7,9,15,21)],
            3:np.arange(1, 150, 20) / 100.
        }
```

```
for ftype in ftype_all.keys():
            # minimaise delta with average spectra for each type
            test_df2 = test_df[test_df.mark == ftype]
            spec_all = test_df2.intence
            aver_spec = np.mean(np.array([np.array(x) for x in spec_all]), axis=0)
            aver_spec_type[ftype] = aver_spec
            data_smooth1[ftype] = {}
            # smooth
            for t in smooth_type.keys():
                print ftype, 'smooth type: ', t,
                data_smooth1[ftype][t] = {}
                if t != 2:
                    data_smooth1[ftype][t]['no'] = {}
                    for p in smooth_type[t]:
                        s = []
                        for spec_i in test_df2.intence:
                            s.append(np.array(smooth(spec_i, [p], type_smooth=t)))
                        now_mean = np.mean(np.array([np.array(x) for x in s]), axis=0)
                        delta = (aver_spec - now_mean) / np.average(aver_spec) * 100.
                        data_smooth1[ftype][t]['no'][p] = np.mean(delta[:2500])
                else:
                    data_smooth1[ftype][t] = {}
                    for p in smooth_type[t][0]:
                        data_smooth1[ftype][t][p] = {}
                        for p2 in smooth_type[t][1]:
                            s = \prod
                            for spec_i in test_df2.intence:
                                s.append(np.array(smooth(spec_i, [p, p2], type_smooth=t)))
                            now_mean = np.mean(np.array([np.array(x) for x in s]), axis=0)
                            delta = (aver_spec - now_mean)/ np.average(aver_spec) * 100.
                            data_smooth1[ftype][t][p][p2] = np.mean(delta[:2500])
                print 'done'
        print 'All Done'
Counter({'4.30.15.16': 24})
4.30.15.16 smooth type: 0 done
4.30.15.16 smooth type: 1 done
4.30.15.16 smooth type: 2 done
4.30.15.16 smooth type: 3 done
All Done
In [7]: # aver_spec_type - include average spec for first key
        mpl.rcParams.update({'font.size': 14})
        for f in data_smooth1:
```

data_smooth1 = {}

```
# Create structure for plot
counter = 1
metrik_data=[]
x_{data} = []
metrik_name = []
for a in data_smooth1[f]:
    if a in [0,1,3]:
        temp_d = []
        temp_x = []
        for p2 in data_smooth1[f][a]['no']:
            temp_d.append(data_smooth1[f][a]['no'][p2])
            temp_x.append(p2)
        temp_x, temp_d = (list(x) for x in zip(*sorted(zip(temp_x, temp_d), key=lamb
        metrik_data.append(temp_d)
        x_data.append(temp_x)
        if a == 1:
            metrik_name.append(str(counter) + u'. Average')
        elif a == 0:
            metrik_name.append(str(counter) + u'. Median')
        else:
            metrik_name.append(str(counter) + u'. Fourier')
        counter += 1
    else:
        for p1 in data_smooth1[f][a]:
            temp_d = []
            temp_x = []
            for p2 in data_smooth1[f][a][p1]:
                temp_d.append(data_smooth1[f][a][p1][p2])
                temp_x.append(p2)
            temp_x, temp_d = (list(x) for x in zip(*sorted(zip(temp_x, temp_d), key=
            metrik_data.append(temp_d)
            x_data.append(temp_x)
            metrik_name.append(str(counter) + u'. SG ord.='+str(p1))
            counter += 1
# create structure for two axis
metrik_data1, metrik_data2 = [], []
x_{data1}, x_{data2} = [], []
metrik_name1, metrik_name2 = [], []
for i in xrange(len(metrik_data)):
    if u'SG' in metrik_name[i] or u'. Average' in metrik_name[i]:
        metrik_data1.append(metrik_data[i])
        x_data1.append(x_data[i])
        metrik_name1.append(metrik_name[i])
    else:
        metrik_data2.append(metrik_data[i])
        x_data2.append(x_data[i])
        metrik_name2.append(metrik_name[i])
```

```
c=0
fig, ax1 = plt.subplots()
ax2 = ax1.twinx()
for i in xrange(len(metrik_data1)):
   y = metrik_data1[i]
    x = x_data1[i]
    ax2.plot(x, y, markers_line[c], label=metrik_name1[i], c=colors[i])
    ax2.text(x[-1], y[-1], metrik_name1[i].split('.')[0],
          horizontalalignment='center',
          bbox=dict(alpha=.5, edgecolor='w', facecolor='w'))
    c+=1
for j in xrange(len(metrik_data2)):
    y = metrik_data2[j]
    x = x_{data2[j]}
    ax1.plot(x, y, markers_line[c], label=metrik_name2[j], c=colors[j+len(metrik_name
    ax1.text(x[-1], y[-1], metrik_name2[j].split('.')[0],
          horizontalalignment='center',
          bbox=dict(alpha=.5, edgecolor='w', facecolor='w'))
    c+=1
ax1.set_xlabel(u'Window Width, channels')
ax1.set_ylabel(u'Delta With Average, %')
ax1.legend(loc='upper left')
ax2.legend(loc='upper right')
plt.draw()
plt.savefig(f+' spec_smooth_aver_eng.png', dpi=300)
plt.show()
```



In [8]: # std-optimization

```
test_df = spe_df[spe_df.fraction == 'grain']
ftype_all = Counter(test_df.ftype)
print ftype_all
aver_spec_type = {}
smooth_type = {
    0:xrange(2,23,5),
    1:xrange(2,23,5),
    2: [xrange(2,6), (7,9,15,21)],
    3:np.arange(1, 150, 20) / 100.
}
data_smooth2 = {}
for ftype in ftype_all.keys():
    # minimaise delta with average spectra for each type
    test_df2 = test_df[test_df.ftype == ftype]
    spec_all = test_df2.intence
    aver_std = np.std(np.array([np.array(x) for x in spec_all]), axis=0)
    aver_spec_type[ftype] = aver_std
    data_smooth2[ftype] = {}
    print np.average(aver_std)
```

```
# smooth
            for t in smooth_type.keys():
                print ftype, 'smooth type: ', t,
                data_smooth2[ftype][t] = {}
                if t != 2:
                    data_smooth2[ftype][t]['no'] = {}
                    for p in smooth_type[t]:
                        s = []
                        for spec_i in test_df2.intence:
                             s.append(np.array(smooth(spec_i, [p], type_smooth=t)))
                        std_spec_i = np.std(s, axis=0)
                        data_smooth2[ftype][t]['no'][p] = np.average(std_spec_i, axis=0) / np.av
                else:
                    data_smooth2[ftype][t] = {}
                    for p in smooth_type[t][0]:
                        data_smooth2[ftype][t][p] = {}
                        for p2 in smooth_type[t][1]:
                            s = []
                            for spec_i in test_df2.intence:
                                 s.append(np.array(smooth(spec_i, [p,p2], type_smooth=t)))
                            data_smooth2[ftype][t][p][p2] = np.average(np.std(s, axis=0),axis=0)
                print 'done'
        print 'All Done'
Counter({'npks': 24})
19.621613426299703
npks smooth type: 0 done
npks smooth type: 1 done
npks smooth type:
                   2 done
npks smooth type: 3 done
All Done
In [9]: mpl.rcParams.update({'font.size': 14})
        for f in data_smooth2:
            # Create structure for plot
            counter = 1
            metrik_data=[]
            x_{data} = []
            metrik_name = []
            for a in data_smooth2[f]:
                if a in [0,1,3]:
                    temp_d = []
                    temp_x = []
                    for p2 in data_smooth2[f][a]['no']:
                        temp_d.append(data_smooth2[f][a]['no'][p2])
                        temp_x.append(p2)
```

```
temp_x, temp_d = (list(x) for x in zip(*sorted(zip(temp_x, temp_d), key=lamb
        metrik_data.append(temp_d)
        x_data.append(temp_x)
        if a == 1:
            metrik_name.append(str(counter) + u'. Average')
        elif a == 0:
            metrik_name.append(str(counter) + u'. Median')
        else:
            metrik_name.append(str(counter) + u'. Fourier')
        counter += 1
    else:
        for p1 in data_smooth2[f][a]:
            temp_d = []
            temp_x = []
            for p2 in data_smooth2[f][a][p1]:
                temp_d.append(data_smooth2[f][a][p1][p2])
                temp_x.append(p2)
            temp_x, temp_d = (list(x) for x in zip(*sorted(zip(temp_x, temp_d), key=
            metrik_data.append(temp_d)
            x_data.append(temp_x)
            metrik_name.append(str(counter) + u'. SG or.='+str(p1))
            counter += 1
# create structure for two axis
metrik_data1, metrik_data2 = [], []
x_{data1}, x_{data2} = [], []
metrik_name1, metrik_name2 = [], []
for i in xrange(len(metrik_data)):
    if u'' in metrik_name[i] or u'. Average' in metrik_name[i]:
        metrik_data1.append(metrik_data[i])
        x_data1.append(x_data[i])
        metrik_name1.append(metrik_name[i])
    else:
        metrik_data2.append(metrik_data[i])
        x_data2.append(x_data[i])
        metrik_name2.append(metrik_name[i])
c=0
fig, ax1 = plt.subplots()
ax2 = ax1.twinx()
for i in xrange(len(metrik_data1)):
    y = metrik_data1[i]
    x = x_data1[i]
    ax2.plot(x, y, markers_line[c], label=metrik_name1[i], c=colors[i])
    ax2.text(x[-1], y[-1], metrik_name1[i].split('.')[0],
          horizontalalignment='center',
          bbox=dict(alpha=.5, edgecolor='w', facecolor='w'))
```

```
c += 1
      for j in xrange(len(metrik_data2)):
          y = metrik_data2[j]
          x = x_{data2[j]}
          ax1.plot(x, y, markers_line[c], label=metrik_name2[j], c=colors[j+len(metrik_name
          ax1.text(x[-1], y[-1], metrik_name2[j].split('.')[0],
                horizontalalignment='center',
                bbox=dict(alpha=.5, edgecolor='w', facecolor='w'))
          c+=1
      ax1.set_xlabel(u'Window Width, channels')
      ax1.set_ylabel(u'STD, %')
      ax1.legend(loc='upper left')
      ax2.legend(loc='upper right')
      plt.draw()
      plt.savefig(f+'_spec_smooth_std_eng.png', dpi=300)
      plt.show()
                                                                      92
    100

    Median

                                                      Average
                    SG or.=2
                                                                      90
     95
                    $G or.<u>≡</u>3
                                                                      88
                  5. SG or.=4
     90
%

 SG or. ≠5

                                                                      86
                  7. Fourier
                                                               6
     85
                                                                      84
     80
                                                                      82
     75
                                   10
                                                15
          0
                       5
                                                            20
```

```
In [10]: # max_peak-optimization

# smoothing spectra

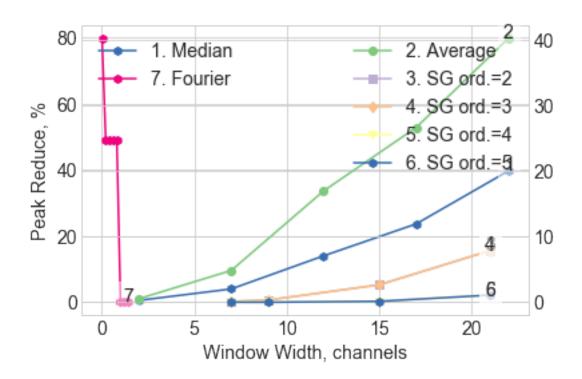
test_df = spe_df[spe_df.fraction == 'grain']
ftype_all = Counter(test_df.ftype)
```

Window Width, channels

```
print ftype_all
         aver_spec_type = {}
         smooth_type = {
             0: xrange(2,23,5),
             1:xrange(2,23,5),
             2: [xrange(2,6), (7,9,15,21)],
             3:np.arange(1, 150, 20) / 100.
         }
         data_smooth3 = {}
         for ftype in ftype_all.keys():
             # minimaise delta with average spectra for each type
             test_df2 = test_df[test_df.ftype == ftype]
             spec_all = test_df2.intence
             aver_spec = np.mean(np.array([np.array(x) for x in spec_all]), axis=0)
             aver_spec_type[ftype] = aver_spec
             data_smooth3[ftype] = {}
             # smooth
             for t in smooth_type.keys():
                 print ftype, 'smooth type: ', t,
                 data_smooth3[ftype][t] = {}
                 if t != 2:
                     data_smooth3[ftype][t]['no'] = {}
                     for p in smooth_type[t]:
                         s = []
                         for spec_i in test_df2.intence:
                             s.append(np.max(smooth(spec_i, [p], type_smooth=t)))
                         std_spec_i = np.average(s)
                         data_smooth3[ftype][t]['no'][p] = - (std_spec_i - np.max(aver_spec)) /
                 else:
                     data_smooth3[ftype][t] = {}
                     for p in smooth_type[t][0]:
                         data_smooth3[ftype][t][p] = {}
                         for p2 in smooth_type[t][1]:
                             for spec_i in test_df2.intence:
                                  s.append(np.max(smooth(spec_i, [p,p2], type_smooth=t)))
                             std_spec_i = np.average(s)
                             data_smooth3[ftype][t][p][p2] = - (std_spec_i - np.max(aver_spec))
                 print 'done'
         print 'All Done'
Counter({'npks': 24})
npks smooth type: 0 done
npks smooth type: 1 done
```

```
npks smooth type: 2 done
npks smooth type: 3 done
All Done
In [11]: mpl.rcParams.update({'font.size': 14})
         for f in data_smooth3:
             # Create structure for plot
             counter = 1
             metrik_data=[]
             x_{data} = []
             metrik_name = []
             for a in data_smooth3[f]:
                 if a in [0,1,3]:
                     temp_d = []
                     temp_x = []
                     for p2 in data_smooth3[f][a]['no']:
                         temp_d.append(data_smooth3[f][a]['no'][p2])
                         temp_x.append(p2)
                     temp_x, temp_d = (list(x) for x in zip(*sorted(zip(temp_x, temp_d), key=lam
                     metrik_data.append(temp_d)
                     x_data.append(temp_x)
                     if a == 1:
                         metrik_name.append(str(counter) + u'. Average')
                     elif a == 0:
                         metrik_name.append(str(counter) + u'. Median')
                     else:
                         metrik_name.append(str(counter) + u'. Fourier')
                     counter += 1
                 else:
                     for p1 in data_smooth3[f][a]:
                         temp_d = []
                         temp_x = []
                         for p2 in data_smooth3[f][a][p1]:
                             temp_d.append(data_smooth3[f][a][p1][p2])
                             temp_x.append(p2)
                         temp_x, temp_d = (list(x) for x in zip(*sorted(zip(temp_x, temp_d), key
                         metrik_data.append(temp_d)
                         x_data.append(temp_x)
                         metrik_name.append(str(counter) + u'. SG ord.='+str(p1))
                         counter += 1
             # create structure for two axis
             metrik_data1, metrik_data2 = [], []
             x_{data1}, x_{data2} = [], []
             metrik_name1, metrik_name2 = [], []
             for i in xrange(len(metrik_data)):
```

```
if u'SG' in metrik_name[i] or u'. Average' in metrik_name[i]:
        metrik_data1.append(metrik_data[i])
        x_data1.append(x_data[i])
        metrik_name1.append(metrik_name[i])
    else:
        metrik_data2.append(metrik_data[i])
        x_data2.append(x_data[i])
        metrik_name2.append(metrik_name[i])
fig, ax1 = plt.subplots()
ax2 = ax1.twinx()
for i in xrange(len(metrik_data1)):
    y = metrik_data1[i]
    x = x_data1[i]
    ax2.plot(x, y, markers_line[i], label=metrik_name1[i], c=colors[i])
    ax2.text(x[-1], y[-1], metrik_name1[i].split('.')[0],
          horizontalalignment='center',
          bbox=dict(alpha=.5, edgecolor='w', facecolor='w'))
for j in xrange(len(metrik_data2)):
    y = metrik_data2[j]
    x = x_data2[i]
    ax1.plot(x, y, markers_line[i], label=metrik_name2[j], c=colors[j+len(metrik_name2[j])]
    ax1.text(x[-1], y[-1], metrik_name2[j].split('.')[0],
          horizontalalignment='center',
          bbox=dict(alpha=.5, edgecolor='w', facecolor='w'))
ax1.set_xlabel(u'Window Width, channels')
ax1.set_ylabel(u'Peak Reduce, %')
ax1.legend(loc='upper left')
ax2.legend(loc='upper right')
plt.draw()
plt.savefig(f+'_spec_smooth_max_eng.png', dpi=300)
plt.show()
```



```
In [12]: # shift max_peak-optimization
         # smoothing spectra
         test_df = spe_df[spe_df.fraction == 'grain']
         ftype_all = Counter(test_df.ftype)
         print ftype_all
         aver_spec_type = {}
         smooth_type = {
             0: xrange(2, 23, 5),
             1:xrange(2,23,5),
             2: [xrange(2,6), (7,9,15,21)],
             3:np.arange(1, 150, 20) / 100.
         }
         data_smooth4 = {}
         for ftype in ftype_all.keys():
             # minimaise delta with average spectra for each type
             test_df2 = test_df[test_df.ftype == ftype]
             spec_all = test_df2.intence
             aver_spec = np.mean(np.array([np.array(x) for x in spec_all]), axis=0)
             aver_spec_type[ftype] = np.where(aver_spec == np.max(aver_spec))[0][0]
             data_smooth4[ftype] = {}
```

```
# smooth
             for t in smooth_type.keys():
                 print ftype, 'smooth type: ', t,
                 data_smooth4[ftype][t] = {}
                 if t != 2:
                     data_smooth4[ftype][t]['no'] = {}
                     for p in smooth_type[t]:
                         s = []
                         for spec_i in test_df2.intence:
                             ss = smooth(spec_i, [p], type_smooth=t)
                             s.append(np.where(ss == np.max(ss))[0][0])
                         std_spec_i = (aver_spec_type[ftype] - np.abs(np.average(s)))/ aver_spec
                         data_smooth4[ftype][t]['no'][p] = std_spec_i
                 else:
                     data_smooth4[ftype][t] = {}
                     for p in smooth_type[t][0]:
                         data_smooth4[ftype][t][p] = {}
                         for p2 in smooth_type[t][1]:
                             s = []
                             for spec_i in test_df2.intence:
                                 ss = smooth(spec_i, [p,p2], type_smooth=t)
                                 s.append(np.where(ss == np.max(ss))[0][0])
                             std_spec_i = (aver_spec_type[ftype] - np.abs(np.average(s)))/ aver_
                             data_smooth4[ftype][t][p][p2] = std_spec_i
                 print 'done'
         print 'All Done'
Counter({'npks': 24})
npks smooth type: 0 done
npks smooth type: 1 done
npks smooth type: 2 done
npks smooth type: 3 done
All Done
In [13]: mpl.rcParams.update({'font.size': 14})
         for f in data_smooth4:
             # Create structure for plot
             counter = 1
             metrik_data=[]
             x_{data} = []
             metrik_name = []
             for a in data_smooth4[f]:
                 if a in [0,1,3]:
                     temp_d = []
                     temp_x = []
```

```
for p2 in data_smooth4[f][a]['no']:
            temp_d.append(data_smooth4[f][a]['no'][p2])
            temp_x.append(p2)
        temp_x, temp_d = (list(x) for x in zip(*sorted(zip(temp_x, temp_d), key=lam
        metrik_data.append(temp_d)
        x_data.append(temp_x)
        if a == 1:
            metrik_name.append(str(counter) + u'. Average')
        elif a == 0:
            metrik_name.append(str(counter) + u'. Median')
        else:
            metrik_name.append(str(counter) + u'. Fourier')
        counter += 1
    else:
        for p1 in data_smooth4[f][a]:
            temp_d = []
            temp_x = []
            for p2 in data_smooth4[f][a][p1]:
                temp_d.append(data_smooth4[f][a][p1][p2])
                temp_x.append(p2)
            temp_x, temp_d = (list(x) for x in zip(*sorted(zip(temp_x, temp_d), key
            metrik_data.append(temp_d)
            x_data.append(temp_x)
            metrik_name.append(str(counter) + u'. SG ord.='+str(p1))
            counter += 1
# create structure for two axis
metrik_data1, metrik_data2 = [], []
x_{data1}, x_{data2} = [], []
metrik_name1, metrik_name2 = [], []
for i in xrange(len(metrik_data)):
    if u'SG' in metrik_name[i] or u'. Average' in metrik_name[i]:
        metrik_data1.append(metrik_data[i])
        x_data1.append(x_data[i])
        metrik_name1.append(metrik_name[i])
    else:
        metrik_data2.append(metrik_data[i])
        x_data2.append(x_data[i])
        metrik_name2.append(metrik_name[i])
fig, ax1 = plt.subplots()
ax2 = ax1.twinx()
for i in xrange(len(metrik_data1)):
   y = metrik_data1[i]
    x = x_data1[i]
    ax2.plot(x, y, markers_line[i], label=metrik_name1[i], c=colors[i])
    ax2.text(x[-1], y[-1], metrik_name1[i].split('.')[0],
          horizontalalignment='center',
```

```
bbox=dict(alpha=.5, edgecolor='w', facecolor='w'))
        for j in xrange(len(metrik_data2)):
            y = metrik_data2[j]
            x = x_{data2[j]}
            ax1.plot(x, y, markers_line[i], label=metrik_name2[j], c=colors[j+len(metrik_name2[j])]
            ax1.text(x[-1], y[-1], metrik_name2[j].split('.')[0],
                  horizontalalignment='center',
                  bbox=dict(alpha=.5, edgecolor='w', facecolor='w'))
        ax1.set_xlabel(u'Window Width, channels')
        ax1.set_ylabel(u'Peak Shift, %')
        ax1.legend(loc='upper left')
        ax2.legend(loc='lower right')
        plt.draw()
        plt.savefig(f+'_spec_smooth_shift_eng.png', dpi=300)
        plt.show()

    Median

                   Fourier
     0
                                                                    -0.05
Peak Shift, %
                                                                    -0.10
    -2
                                                  2. Average
                                                                    -0.15
                                                  3. SG ord.=2
                                                  4. SG ord.=3
    -4
                                                                    -0.20
                                                  5. SG ord.=4
                                                  6. SG ord.=5
                                                                    -0.25
                      5
                                 10
                                              15
                                                          20
                       Window Width, channels
```

```
In [14]: # Find of base line

    test_df = spe_df[spe_df.fraction == 'grain']
    ftype_all = Counter(test_df.ftype)

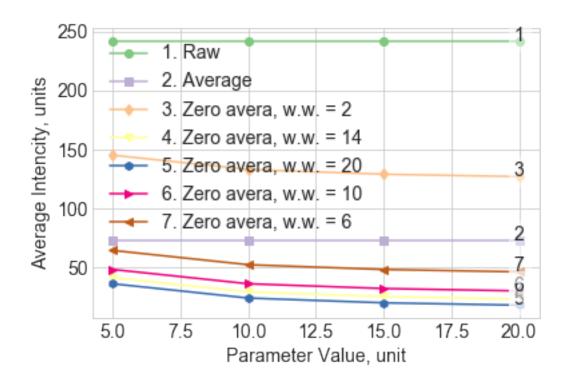
base_type = {
    1:[[2,6,10,14,20], xrange(5,21,5)],
```

```
2:xrange(2,40,5),
    3: xrange(10,40,5) # .. -
}
baseline_spec_type = {}
baseline_spec_aver_int = {}
baseline_spec_aver_len = {}
data_base_max = \{\}
data_base_len = {}
for ftype in ftype_all.keys():
    # minimaise delta with average spectra for each type
    test_df2 = test_df[test_df.ftype == ftype]
    spec_all = test_df2.intence
    all_spec = np.array([np.array(x) for x in spec_all])
    aver_spec = np.mean(all_spec, axis=0)
    baseline_spec_type[ftype] = aver_spec
    baseline_spec_aver_int[ftype] = np.average(aver_spec[aver_spec < np.mean(aver_spec)
    baseline_spec_aver_len[ftype] = len(aver_spec[aver_spec < np.mean(aver_spec)])</pre>
    data_base_max[ftype] = {}
    data_base_len[ftype] = {}
    # find baseline
    for t in base_type.keys():
        print ftype, 'baseline type: ', t,
        data_base_max[ftype][t] = {}
        data_base_len[ftype][t] = {}
        if t != 1:
            data_base_max[ftype][t]['no'] = {}
            data_base_len[ftype][t]['no'] = {}
            for p in base_type[t]:
                s = []
                len_s = []
                for spec_i in test_df2.intence:
                    ss = zero_area(spec_i, [p], type_alg=t)
                    len_s.append(np.average(np.array(spec_i) - np.array(ss))) # delta
                    s.append(np.average(ss)) # average
                data_base_max[ftype][t]['no'][p] = np.average(s)
                data_base_len[ftype][t]['no'][p] = np.average(len_s)
        else:
            for p in base_type[t][0]:
                print p,
                data_base_max[ftype][t][p]={}
                data_base_len[ftype][t][p]={}
                for p2 in base_type[t][1]:
                    print ' -', p2,
                    s = []
                    len_s = []
                    for spec_i in test_df2.intence:
```

```
ss = zero_area(spec_i, [p, p2], type_alg=t)
                                  len_s.append(np.average(np.array(spec_i) - np.array(ss)))
                                  s.append(np.average(ss))
                              data_base_max[ftype][t][p][p2]=np.average(s)
                              data_base_len[ftype][t][p][p2]=np.average(len_s)
                         print ' +'
                 print 'done'
         print 'All Done'
npks baseline type: 1 2 - 5 - 10 - 15 - 20 +
6 - 5 - 10 - 15 - 20 +
10 - 5 - 10 - 15 - 20 +
14 - 5 - 10 - 15 - 20 +
20 - 5 - 10 - 15 - 20 +
done
npks baseline type: 2 done
npks baseline type: 3
/usr/local/lib/python2.7/dist-packages/ipykernel_launcher.py:248: RuntimeWarning: invalid value
 done
All Done
In [15]: mpl.rcParams.update({'font.size': 14})
         for f in data_base_max:
         # Create structure for plot
             print f
             spec_aver_int = np.mean(baseline_spec_type[f])
             spec_aver_len = len(baseline_spec_type[f])
             sver_filter_int = np.mean(baseline_spec_type[f][baseline_spec_type[f] < np.mean(baseline_spec_type[f])</pre>
             sver_filter_len = len(baseline_spec_type[f][baseline_spec_type[f] < np.mean(baseline_spec_type[f])</pre>
             counter = 3
             metrik_data_max = [[spec_aver_int]*4, [sver_filter_int]*4]
             x_{data} = [range(5,21,5), range(5,21,5)]
             metrik_data_len = [[spec_aver_len]*4, [sver_filter_len]*4]
             metrik_name = [u'1. Raw', u'2. Average']
             for a in data_base_max[f]:
                 if a in [2,3]:
                     temp_d_max = []
                     temp_d_len = []
                     temp_x = []
                     for p2 in data_base_max[f][a]['no']:
                         temp_d_max.append(data_base_max[f][a]['no'][p2])
                         temp_d_len.append(data_base_len[f][a]['no'][p2])
                         temp_x.append(p2)
```

```
_, temp_d_max = (list(x) for x in zip(*sorted(zip(temp_x, temp_d_max), key=
        temp_x, temp_d_len = (list(x) for x in zip(*sorted(zip(temp_x, temp_d_len),
        metrik_data_max.append(temp_d_max)
        metrik_data_len.append(temp_d_len)
        x_data.append(temp_x)
        if a == 2:
            metrik_name.append(str(counter) + u'. Average Windows')
        else:
            metrik_name.append(str(counter) + u'. Parabola')
        counter += 1
    else:
        for p1 in data_base_max[f][a]:
            temp_d_max = []
            temp_d_len = []
            temp_x = []
            for p2 in data_base_max[f][a][p1]:
                temp_d_max.append(data_base_max[f][a][p1][p2])
                temp_d_len.append(data_base_len[f][a][p1][p2])
                temp_x.append(p2)
            _, temp_d_max = (list(x) for x in zip(*sorted(zip(temp_x, temp_d_max),
            temp_x, temp_d_len = (list(x) for x in zip(*sorted(zip(temp_x, temp_d_l
            metrik_data_max.append(temp_d_max)
            metrik_data_len.append(temp_d_len)
            x_data.append(temp_x)
            metrik_name.append(str(counter) + u'. Zero avera, w.w. = ' + str(p1))
            counter += 1
fig, ax1 = plt.subplots()
for i in xrange(len(metrik_data_max)):
    if metrik_name[i].split('.')[0] not in ['8','9']:
        y = metrik_data_max[i]
        x = x_{data[i]}
        ax1.plot(x, y, markers_line[i], label=metrik_name[i], c=colors[i])
        ax1.text(x[-1], y[-1], metrik_name[i].split('.')[0],
              horizontalalignment='center',
              bbox=dict(alpha=.5, edgecolor='w', facecolor='w'))
ax1.set_xlabel(u'Parameter Value, unit')
ax1.set_ylabel(u'Average Intencity, units')
ax1.legend(loc='upper left')
plt.draw()
plt.savefig(f+'_spec_baseline_max_eng.png', dpi=300)
plt.show()
```

npks



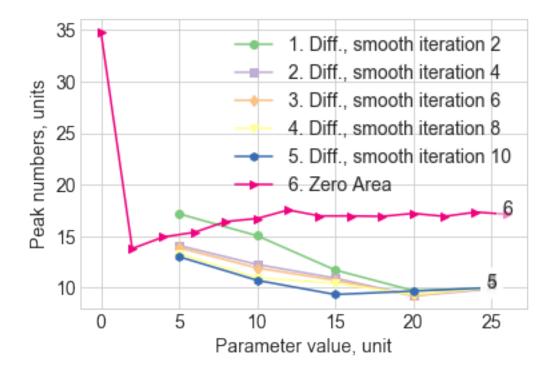
```
In [16]: #
         # Find of base line
         test_df = spe_df[spe_df.fraction == 'grain']
         ftype_all = Counter(test_df.ftype)
         base_type = {
             1: [xrange(0,11,2), xrange(5,26,5)], # iteration of smooth, threshold
             2:xrange(0,27,2) # iteration of smooth
         data_peaks_num = {}
         for ftype in ftype_all.keys():
             # minimaise delta with average spectra for each type
             test_df2 = test_df[test_df.ftype == ftype]
             spec_all = test_df2.intence
             data_peaks_num[ftype] = {}
             # find baseline
             for t in base_type.keys():
                 print ftype, 'baseline type: ', t,
                 data_peaks_num[ftype][t] = {}
                 if t != 1:
                     data_peaks_num[ftype][t]['no'] = {}
                     for p in base_type[t]:
                         s = []
```

```
auto_find = zero_area(spec_i, params=[10,10], type_alg=1)
                             mask_spec = is_baseline(auto_find) # floor=np.average(auto_find[1:]
                             find_peak = calc_peacs(
                                 spec_i,
                                 baseline_mask=mask_spec,
                                 base_fit=calc_baseline(spec_i, mask_spec),
                                 parameters=[p],
                                 type_calc=t
                             [0]
                             s.append(len(find_peak))
                         data_peaks_num[ftype][t]['no'][p] = np.average(s)
                 else:
                     for p in base_type[t][0]:
                         print p,
                         data_peaks_num[ftype][t][p]={}
                         for p2 in base_type[t][1]:
                            print ' -', p2,
                             s = []
                             for spec_i in test_df2.intence:
                                 auto_find = zero_area(spec_i, params=[10,10], type_alg=1)
                                 mask_spec = is_baseline(auto_find) # floor=np.average(auto_find)
                                 find_peak = calc_peacs(
                                    spec_i,
                                    baseline_mask=mask_spec,
                                    base_fit=calc_baseline(spec_i, mask_spec),
                                    parameters=[p, p2],
                                    type_calc=t
                                 [0]
                                 s.append(len(find_peak))
                             data_peaks_num[ftype][t][p][p2]=np.average(s)
                         print ' +'
                print 'done'
        print 'All Done'
npks baseline type: 10 - 5
/usr/local/lib/python2.7/dist-packages/numpy/lib/function_base.py:1128: RuntimeWarning: Mean of
  avg = a.mean(axis)
/usr/local/lib/python2.7/dist-packages/numpy/core/_methods.py:80: RuntimeWarning: invalid value
 ret = ret.dtype.type(ret / rcount)
  - 10 - 15 - 20 - 25 +
2 - 5 - 10 - 15 - 20 - 25 +
4 - 5 - 10 - 15 - 20 - 25 +
6 - 5 - 10 - 15 - 20 - 25 +
8 - 5 - 10 - 15 - 20 - 25 +
```

for spec_i in test_df2.intence:

```
10 - 5 - 10 - 15 - 20 - 25 +
done
npks baseline type: 2 done
All Done
In [17]: mpl.rcParams.update({'font.size': 14})
         for f in data_peaks_num:
         # Create structure for plot
             print f
             counter = 1
             x_{data} = []
             metrik_data = []
             metrik_name = []
             for a in data_peaks_num[f]:
                 if a != 1:  # zero-area
                     temp_d = []
                     temp_x = []
                     for p2 in data_peaks_num[f][a]['no']:
                         temp_d.append(data_peaks_num[f][a]['no'][p2])
                         temp_x.append(p2)
                     temp_x, temp_d = (list(x) for x in zip(*sorted(zip(temp_x, temp_d), key=lam
                     metrik_data.append(temp_d)
                     x_data.append(temp_x)
                     metrik_name.append(str(counter) + u'. Zero Area')
                     counter += 1
                 else:
                     for p1 in data_peaks_num[f][a]:
                         if max(data_peaks_num[f][a][p1].values()) < 70:</pre>
                             temp_d = []
                             temp_x = []
                             for p2 in data_peaks_num[f][a][p1]:
                                  temp_d.append(data_peaks_num[f][a][p1][p2])
                                  temp_x.append(p2)
                             temp_x, temp_d = (list(x) for x in zip(*sorted(zip(temp_x, temp_d),
                             metrik_data.append(temp_d)
                             x_data.append(temp_x)
                             metrik_name.append(str(counter) + u'. Diff., smooth iteration ' + s
                             counter += 1
                         else:
                             pass
             fig, ax1 = plt.subplots()
             for i in xrange(len(metrik_data)):
                 if metrik_name[i].split('.')[0] not in ['8','9']:
                     y = metrik_data[i]
```

npks



1.2

1.3 Work with optimal parameters and save data

frac_types = list(spe_df_cl_type[spe_df_cl_type.ftype=='npks'].fraction)

```
intence = list(spe_df_cl_type[spe_df_cl_type.ftype=='npks'].intence)
marks = ['4.30.15.16']
specs = {} # {type:[[intenses], [], ...]}
names = []
for i in xrange(len(frac_types)):
    type_obj = 'NPK(S) ' + str(marks[0]) + ' ' + str(frac_types[i])
    if type_obj in specs:
        specs[type_obj] append(intence[i])
    else:
        names.append(type_obj)
        specs[type_obj] = [intence[i]]
aver_specs = {}
for i in specs:
    aver_all = 0
    for j in specs[i]:
        aver_all = aver_all + np.array(j)
    aver_all = aver_all / float(len(specs[i]))
    aver_specs[i] = aver_all
print aver_specs.keys()
aver_specs.pop('NPK(S) 4.30.15.16 100.dry')
aver_specs.pop('NPK(S) 4.30.15.16 500.dry')
reverse_k = {
    'NPK(S) 4.30.15.16 rawgrain': 'NPK(S) 4-30-15(16) granules',
    'NPK(S) 4.30.15.16 grain': 'NPK(S) 4-30-15(16) pressed granules',
    'NPK(S) 4.30.15.16 500': 'NPK(S) 4-30-15(16) pressed 500 mkm',
    'NPK(S) 4.30.15.16 100': 'NPK(S) 4-30-15(16) pressed 100 mkm'
}
aver_specs = dict((reverse_k[key], value) for (key, value) in aver_specs.items())
c = np.arange(len(aver_specs['NPK(S) 4-30-15(16) granules']))
energy = np.average(spe_df_cl_type['a0']) + np.average(spe_df_cl_type['a1'])*c + np.ave
num_plot = len(aver_specs.keys())
fig, ax = plt.subplots(num_plot, figsize=(num_plot*4,num_plot*4))
names_all = list(aver_specs.keys())
print aver_specs.keys()
gauss_h = {}
gauss_s = \{\}
for i in range(num_plot):
```

```
ax[i].plot(energy, fit_baseline, '--', label = u'Approximation of Baseline', linewi
             ax[i].plot(energy[mask_spec], spec[mask_spec], label = u'Selected Baseline')
             my\_xticks = [0]
             find_peak = calc_peacs(spec, baseline_mask=mask_spec, base_fit=fit_baseline, parame
             for j in range(len(find_peak[1])):
                 if find_peak[1][j] < 2500:
                     c = find_peak[1][j]
                     f_e = spe_df['a0'][0] + spe_df['a1'][0]*c + spe_df['a2'][0]*c*c
                     if (round(f_e, 1) not in my_xticks) and (find_peak[0][j] > 2 * find_peak[-1
                         my_xticks.append(round(f_e, 2))
                         gauss_h[i].append(find_peak[2][j])
                         gauss_s[i].append(find_peak[3][j])
                         ax[i].plot([f_e for x in range(int(find_peak[0][j]))],
                                    [x for x in range(int(find_peak[0][j]))], c=colors[-1])
             ax[i].set_xlim((0, 20))
             ax[i].set_ylim((0, 3000))
             ax[i].set_title(names_all[i])
             ax[i].legend(loc = "upper right")
             my_xticks.append(20)
             ax[i].set_xticks(my_xticks)
             ax[i].set_xticklabels(my_xticks, rotation='vertical', fontsize=14)
        ax[-1].set_xlabel(u'Energy, kEv', size=21)
        ax[1].set_ylabel(u'Intencity, imp.', size=21)
        plt.tight_layout()
        plt.draw()
        plt.savefig('aver_spec_npks_fract.png', dpi=300)
        plt.show()
['NPK(S) 4.30.15.16 500.dry', 'NPK(S) 4.30.15.16 rawgrain', 'NPK(S) 4.30.15.16 100.dry', 'NPK(S)
['NPK(S) 4-30-15(16) granules', 'NPK(S) 4-30-15(16) pressed 100 mkm', 'NPK(S) 4-30-15(16) pressed
```

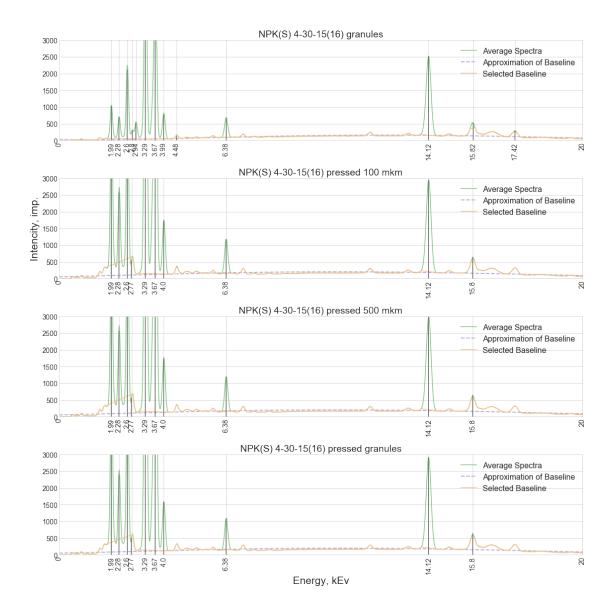
name = names_all[i]
gauss_h[i] = []
gauss_s[i] = []

spec = aver_specs[name]

ax[i].plot(energy, spec, label = u'Average Spectra')

auto_find = zero_area(spec, params=[14,10], type_alg=1)

mask_spec = is_baseline(auto_find) # floor=np.average(auto_find[1:]))
fit_baseline = calc_baseline(spec_i=spec, baseline_mask=mask_spec)



1.4 Prepare all spectra with select algorithm

```
spec = spe_df_cl_type.intence[i]
             print j, '. ', i, ' start... ',
             spec = smooth(spec_i=spec, params=[4, 15], type_smooth=2)
             auto_find = zero_area(spec, params=[10,10], type_alg=1)
             mask_spec = is_baseline(auto_find) #floor=np.average(auto_find[1:]))
             fit_baseline = calc_baseline(spec_i=spec, baseline_mask=mask_spec)
             all_peaks_int = []
             all_peaks_chan = []
             all_peak_g_h = []
             all_peak_g_s = []
             all_peak_b = []
             find_peak = calc_peacs(spec, baseline_mask=mask_spec, base_fit=fit_baseline, parame
             for j in xrange(len(find_peak[1])):
                 if find_peak[1][j] < 2500 and find_peak[1][j] > 2.*find_peak[-1][j]:
                     all_peaks_int.append(find_peak[0][j])
                     all_peaks_chan.append(int(find_peak[1][j]))
                     all_peak_g_h.append(find_peak[2][j])
                     all_peak_g_s.append(find_peak[3][j])
                     all_peak_b.append(find_peak[4][j])
             spe_df_cl_type.set_value(i, 'peaks_int', all_peaks_int)
             spe_df_cl_type.set_value(i, 'peaks_chan', all_peaks_chan)
             spe_df_cl_type.set_value(i, 'gauss_int', all_peak_g_h)
             spe_df_cl_type.set_value(i, 'gauss_square', all_peak_g_s)
             spe_df_cl_type.set_value(i, 'base_peaks_int', all_peak_b)
             spe_df_cl_type.set_value(i, 'base_s', np.sum(fit_baseline[0:2500]))
             base_max = np.max(fit_baseline[0:2500])
             spe_df_cl_type.set_value(i, 'base_max_int', base_max)
             print 'done'
         print 'All Done'
all data: 118
0 . 0 start...
/usr/local/lib/python2.7/dist-packages/ipykernel_launcher.py:35: FutureWarning: set_value is dep
/usr/local/lib/python2.7/dist-packages/ipykernel_launcher.py:36: FutureWarning: set_value is dep
/usr/local/lib/python2.7/dist-packages/ipykernel_launcher.py:38: FutureWarning: set_value is dep
/usr/local/lib/python2.7/dist-packages/ipykernel_launcher.py:39: FutureWarning: set_value is dep
/usr/local/lib/python2.7/dist-packages/ipykernel_launcher.py:40: FutureWarning: set_value is dep
/usr/local/lib/python2.7/dist-packages/ipykernel_launcher.py:42: FutureWarning: set_value is dep
/usr/local/lib/python2.7/dist-packages/ipykernel_launcher.py:45: FutureWarning: set_value is dep
```

for j in xrange(len(names)):

i = names[j]

done 1 . 1 start... done 2 . 2 done start... 3 . 3 start... done 4 start... done 5 5 start... done 6. 6 start... done 7. 7 start... done 8 start... done 9. 9 start... done 10 . 10 start... done 11 . 11 start... done 12 . 12 start... done 13 . 13 start... done 14 . 14 start... done 15 15 done start... 16 . 16 start... done 17 . 17 start... done 18 . 18 start... done 19 19 start... done 20 20 . start... done 21 . 21 start... done 22 . 22 start... done 23 . 23 start... done 24 . 24 start... done 25 . 25 start... done 26 26 start... done 27 . 27 start... done 28 . 28 start... done 29 29 start... done 30 . 30 start... done 31 . 31 start... done 32 . 32 start... done 33 . 33 start... done 34 . 34 start... done 35 . 35 start... done 36 . 36 start... done 37 . 37 start... done 38 . 38 start... done 39 . 39 start... done 40 . 40 start... done 41 . 41 start... done 42 . 42 start... done 43 . 43 start... done 44 . 44 start... done 45 . 45 start... done

46 .

47 .

46

47

start...

start...

done

done

```
48 .
      48
                       done
           start...
49 .
      49
           start...
                       done
50 .
      50
                       done
           start...
51 .
       51
           start...
                       done
52 .
       52
           start...
                       done
53 .
       53
           start...
                       done
54 .
       54
           start...
                       done
55 .
       55
           start...
                       done
56 .
       56
           start...
                       done
57 .
      57
           start...
                       done
58 .
       58
           start...
                       done
59
       59
           start...
                       done
60 .
       60
           start...
                       done
61 .
       61
           start...
                       done
62 .
       62
           start...
                       done
63 .
       63
                       done
           start...
64 .
       64
           start...
                       done
65 .
       65
           start...
                       done
66 .
       66
           start...
                       done
67 .
       67
           start...
                       done
68 .
       68
           start...
                       done
69 .
       69
           start...
                       done
70 .
      70
           start...
                       done
71 .
      71
           start...
                       done
72 .
      72
           start...
                       done
73 .
      73
           start...
                       done
74
      74
           start...
                       done
75 .
      75
           start...
                       done
76 .
      76
           start...
                       done
77 .
      77
           start...
                       done
78 .
       78
           start...
                       done
79 .
       79
           start...
                       done
80 .
      80
           start...
                       done
81 .
      81
           start...
                       done
82 .
       82
           start...
                       done
83 .
       83
           start...
                       done
84 .
       84
           start...
                       done
85 .
      85
           start...
                       done
86 .
       86
           start...
                       done
87 .
      87
           start...
                       done
88 .
      88
                       done
           start...
89 .
       89
           start...
                       done
90 .
       90
           start...
                       done
91 .
       91
           start...
                       done
92 .
       92
           start...
                       done
93 .
       93
           start...
                       done
94 .
       94
           start...
                       done
95 .
      95
           start...
                       done
```

```
96 .
     96 start...
                  done
97 .
     97
                  done
         start...
     98
         start...
                  done
99 .
     99
         start...
                  done
100 . 100 start...
                    done
101 .
      101 start...
                    done
102 .
      102 start...
                    done
103 .
      103 start...
                    done
104 .
      104 start...
                    done
105 .
      105 start...
                    done
106 .
      106 start...
                    done
107 . 107
          start...
                    done
108 .
      108 start...
                    done
109 .
      109 start...
                    done
110 . 110 start...
                    done
111 . 111 start...
                    done
112 . 112 start...
                    done
113 . 113 start...
                    done
114 . 114 start...
                    done
115 . 115 start...
                    done
116 . 116 start...
                    done
117 . 117 start...
                    done
All Done
```

In [20]: # Create elements columns and calculate characteristics line in X-ray spectrum

```
target_peaks = {
    1.7: 'Si',
    2.0: 'P',
    2.3: 'S',
    2.6: 'Cl',
    3.3: 'K',
    3.7: 'Ca',
    4.5: 'Ti',
    5.9: 'Mn',
    6.4: 'Fe',
    8.1: 'Ta',
    8.6: 'Zn',
    14.2: 'Sr',
    17.4: 'Mo',
    16.5: 'Mo_Coh'
}
ind = spe_df_cl_type.index.tolist()
col_name = target_peaks.keys()
# construct data {col_name: index_value}
```

```
data_all = {}
for e in target_peaks:
    data_all[e] = []
    for i in ind:
        c = int((e - spe_df_cl_type['a0'][i])/spe_df_cl_type['a1'][i])
        d = np.average(spe_df_cl_type['intence'][i][c-1:c+1])
        data_all[e].append(d)
spe_df2 = pd.DataFrame(columns=col_name, index=ind, data=data_all) # df with channel or
# energy list (equal of channel list) for each index
energy_all = pd.DataFrame(index=ind, data={'energy':[np.average(spe_df_cl_type['a0']) +
            np.average(spe_df_cl_type['a1'])*np.array(x) + \
            np.average(spe_df_cl_type['a2'])*np.array(x)*np.array(x) for x in spe_df_cl
for i in xrange(len(ind)): # for all index in df
    i_ind = ind[i] # df index
    data_en = energy_all['energy'][i_ind]
    for en_ind in xrange(len(data_en)): # for each energy in energy list
        en = round(data_en[en_ind], 1)
        if en in col_name: # if energy of finding peaks in list of peaks
            spe_df2[en][i_ind] = spe_df_cl_type['peaks_int'][i_ind][en_ind]
new_names_round = [target_peaks[x] for x in spe_df2.columns.tolist()]
spe_df2.columns = new_names_round
print spe_df2.head()
# Gauss int
ind = spe_df_cl_type.index.tolist()
col_name = target_peaks.keys()
def gauss_peaks2(i_all, channel_all, spec_i):
    # calculate Gauss S and I of peaks
    gauss_s = []
    gauss_i = []
    for i in xrange(len(channel_all)):
        c_middle = channel_all[i]
        g_h = 0
        g_s = 0
        w = 15
        y = np.array(spec_i[c_middle-w:c_middle+w]) # gauss fit of points
        x = np.array(xrange(len(y)))
        # correction for weighted arithmetic mean
        mean = np.sum(x * y) / np.sum(y)
        sigma = np.sqrt(np.sum(y * (x - mean)**2) / np.sum(y))
```

```
try:
            popt,pcov = curve_fit(gaus, x, y, p0=[max(y), mean, sigma])
            y_fit = gaus(x, *popt)
            g_h = np.max(y_fit)
            g_s = g_h * (2 * np.pi) ** 0.5 * popt[-1] # gauss square
        except RuntimeError as var:
            g_h = 0
            g_s = 0
        except ValueError as var:
            print var
            print x, y
            g_h = 0
            g_s = 0
        gauss_i.append(g_h)
        gauss_s.append(g_s)
    return np.array(gauss_i), np.array(gauss_s)
# construct data {col_name: index_value} with calculate value from raw spectra
data_all = {}
data_all2 = {}
for e in target_peaks:
    data_all[e] = []
    data_all2[e] = []
    for i in ind:
        c = int((e - spe_df_cl_type['a0'][i])/spe_df_cl_type['a1'][i])
        spec = spe_df_cl_type['intence'][i]
        d = gauss_peaks2([spec[c]], [c], spec)
        data_all[e].append(d[0][0])
        data_all2[e].append(d[1][0])
spe_df3 = pd.DataFrame(columns=col_name, index=ind, data=data_all) # df with channel of
spe_df4 = pd.DataFrame(columns=col_name, index=ind, data=data_all2) # df with channel
energy_all = pd.DataFrame(index=ind, data={'energy':[np.average(spe_df_cl_type['a0']) +
            np.average(spe_df_cl_type['a1'])*np.array(x) + \
            np.average(spe_df_cl_type['a2'])*np.array(x)*np.array(x) for x in spe_df_cl
for i in xrange(len(ind)): # for all index in df
    i_ind = ind[i] # df index
    data_en = energy_all['energy'][i_ind]
    for en_ind in xrange(len(data_en)): # for each energy in energy list
        en = round(data_en[en_ind], 1)
        if en in col_name:
            spe_df3[en][i_ind] = spe_df_cl_type['gauss_int'][i_ind][en_ind]
new_names_round = ['Gauss.int_' + target_peaks[x] for x in spe_df3.columns.tolist()]
```

```
# spe_df3 = spe_df3.loc[:,~spe_df3.columns.duplicated()]
         print spe_df3.head()
         # Gauss square
         for i in xrange(len(ind)): # for all index in df
            i_ind = ind[i] # df index
            data_en = energy_all['energy'][i_ind]
            for en_ind in xrange(len(data_en)): # for each energy in energy list
                en = round(data_en[en_ind], 1)
                if en in col_name:
                    spe_df4[en][i_ind] = spe_df_cl_type['gauss_square'][i_ind][en_ind]
         new_names_round = ['Gauss.sqe_' + target_peaks[x] for x in spe_df4.columns.tolist()]
         spe_df4.columns = new_names_round
         # spe_df3 = spe_df3.loc[:,~spe_df3.columns.duplicated()]
        print spe_df4.head()
         # Concatenate
         _spe_backup = spe_df_cl_type.copy()
         spe_df_cl_type = pd.concat([spe_df_cl_type, spe_df2], axis=1)
         spe_df_cl_type = pd.concat([spe_df_cl_type, spe_df3], axis=1)
         spe_df_cl_type = pd.concat([spe_df_cl_type, spe_df4], axis=1)
         print spe_df_cl_type.shape
         spe_df_cl_type = spe_df_cl_type.loc[:,~spe_df_cl_type.columns.duplicated()]
        print spe_df_cl_type.shape
          Ti Mo_Coh
                                Ρ
                                     Si
                                            Ta
                                                   Zn
                                                               Fe
                                                                  \
0 129.384529
               272.0 1006.501009 44.0
                                         111.0 101.0 619.249155
1 137.037628
               247.5
                       995.017825 46.0
                                         122.0 105.0 637.019248
2 111.441463
               262.0 905.664878 40.0
                                         104.0 109.5 591.678159
3 182.964675
               274.5 1040.881217
                                   52.5
                                         120.5 101.5 747.982530
4 143.055810
               287.0 1094.369807 49.0
                                         118.0 117.5 730.648810
                                           \operatorname{\mathtt{Sr}}
                                                        Ca
           S
                        K
                                   Mo
                                                               Mn
                           282.513319 1835.0
0 671.294053 6031.282690
                                               4922.477767
                                                           140.5
1 649.770450 5944.970328
                           304.820598 1811.0
                                               4732.075199 119.5
2 677.128804 6254.167335
                           277.907663 1803.0
                                               4785.358494 110.0
3 734.601407 6654.539539
                           315.945697 2057.5
                                               5476.197952 158.0
4 758.327250 6515.351330 300.213075 2077.0 5635.215976 146.0
           C1
0 1940.049791
```

spe_df3.columns = new_names_round

```
2056.749509
2
  2323.368930
3
  2168.094544
  2001.890261
   Gauss.int_Ti
                 Gauss.int_Mo_Coh
                                    Gauss.int_P
                                                  Gauss.int_Si
                                                                 Gauss.int_Ta
     143.242691
                        265.825544
                                     1016.915821
                                                            0.0
                                                                     0.00000
0
1
     146.487391
                        261.039100
                                     1006.491737
                                                            0.0
                                                                   119.916368
     138.061561
2
                        252.316906
                                      911.764661
                                                            0.0
                                                                   114.079939
3
     176.220104
                                    1050.789326
                                                            0.0
                          0.000000
                                                                     0.00000
4
     160.357539
                        278.652902
                                     1108.443352
                                                            0.0
                                                                   122.784574
   Gauss.int_Zn
                 Gauss.int_Fe
                                Gauss.int_S
                                              Gauss.int_K
                                                            Gauss.int_Mo
0
       0.00000
                    617.833272
                                 684.311373
                                              6205.358513
                                                              262.910500
1
                                 654.506083
                                                              273.956953
     114.577794
                    637.824446
                                              6129.051296
2
     110.590734
                    598.111303
                                 678.914034
                                              6435.808649
                                                              265.395286
3
                    755.271740
                                 746.349434
                                              6854.273488
                                                              294.630719
       0.000000
4
     120.006867
                    719.739657
                                 770.405527
                                              6725.137436
                                                              289.095685
                 Gauss.int_Ca
                                Gauss.int_Mn
   Gauss.int_Sr
                                               Gauss.int_Cl
0
    2423.743710
                   4935.508225
                                  136.599797
                                                2023.531240
1
    2395.710670
                   4743.320123
                                  132.595486
                                                2152.406120
    2378.925808
                   4777.895624
                                  126.534640
                                                2430.113743
3
    2718.565583
                   5472.205036
                                  152.124133
                                                2265.172763
    2680.173291
                   5652.474968
                                  144.900317
                                                2087.741793
   Gauss.sqe_Ti
                 Gauss.sqe_Mo_Coh
                                                   Gauss.sqe_Si
                                                                  Gauss.sqe_Ta
                                     Gauss.sqe_P
    3489.316944
                                                                      0.00000
0
                      19457.642251
                                     14137.471208
                                                             0.0
    3537.385776
                                     13797.720353
1
                      17659.154834
                                                             0.0
                                                                   8565.473145
2
    3295.938904
                      22075.067717
                                     12513.218346
                                                             0.0
                                                                  14826.830173
3
                                     14735.362429
    3857.559135
                          0.000000
                                                             0.0
                                                                      0.000000
4
    3808.544581
                      24642.611208
                                     15229.513335
                                                             0.0
                                                                   9478.980972
   Gauss.sqe_Zn
                 Gauss.sqe_Fe
                                 Gauss.sqe_S
                                                Gauss.sqe_K Gauss.sqe_Mo
0
       0.000000
                 12782.834144
                                 9950.664247
                                               86934.671678
                                                              10943.905700
   24134.269235
                 12718.985814
                                 9686.542485
                                               85417.271458
                                                              10321.926453
1
2
   10936.573898
                  12344.963302
                                10116.839276
                                               89172.181946
                                                              11689.414083
3
       0.000000
                  15003.922805
                                10903.042334
                                               95391.063701
                                                              11901.029557
   31863.784002
                  14336.032861
                                10985.610374
                                               93685.496143
                                                              11950.637694
   Gauss.sqe_Sr
                 Gauss.sqe_Ca
                                Gauss.sqe_Mn
                                               Gauss.sqe_Cl
  66125.880105
                 81113.835877
                                 4909.040539
                                               27369.833763
  65017.214651
                 78708.054755
                                               28551.141260
1
                                 5373.082037
 65654.177592
                 79611.013286
                                 4916.262922
                                               32451.903800
  74405.676539
                 89784.619791
                                 5521.977542
                                               30136.133442
4 74258.461860
                                 5885.501915
                 92481.740876
                                               27987.210135
(118, 64)
(118, 64)
```

```
In [21]: spe_df_cl_type.to_pickle('spe_df_cl_type') # save data frame
In [22]: spe_df_cl_type
Out [22]:
                     a1
                         fraction
                                     kd
         0
              0.008923
                         rawgrain
                                   None
         1
              0.008923
                         rawgrain
                                   None
         2
              0.008923 rawgrain
                                   None
         3
              0.008923 rawgrain
                                   None
         4
              0.008923 rawgrain None
         5
              0.008923 rawgrain
                                   None
         6
              0.008923 rawgrain
                                   None
         7
              0.008923 rawgrain
                                   None
         8
              0.008923 rawgrain
                                   None
         9
              0.008923
                         rawgrain
                                   None
         10
              0.008923
                            grain
                                   None
         11
              0.008923
                            grain
                                   None
         12
              0.008923
                            grain
                                   None
         13
              0.008923
                            grain
                                   None
         14
              0.008923
                            grain
                                   None
         15
              0.008923
                            grain
                                   None
         16
              0.008923
                            grain
                                   None
         17
              0.008923
                            grain
                                   None
         18
              0.008923
                            grain
                                   None
         19
              0.008923
                            grain None
         20
              0.008923
                            grain None
         21
              0.008923
                            grain None
         22
              0.008923
                            grain None
         23
              0.008923
                            grain
                                   None
         24
              0.008923
                            grain
                                   None
         25
              0.008923
                            grain
                                   None
         26
              0.008923
                                   None
                            grain
         27
              0.008923
                            grain
                                   None
         28
              0.008923
                            grain
                                   None
         29
              0.008923
                            grain
                                   None
                              . . .
         . .
         88
              0.008923
                              500
                                   None
         89
              0.008923
                          500.dry
                                   None
                          500.dry
         90
              0.008923
                                   None
         91
              0.008923
                          500.dry
                                   None
         92
              0.008923
                          500.dry
                                   None
         93
              0.008923
                          500.dry
                                   None
         94
              0.008923
                          500.dry
                                   None
         95
              0.008923
                          500.dry
                                   None
         96
              0.008923
                          500.drv
                                   None
         97
              0.008923
                          500.dry
                                   None
         98
              0.008923
                          500.dry
                                   None
         99
              0.008923
                          500.dry None
```

```
0.008923
                500.dry
100
                          None
101
     0.008923
                500.dry
                          None
                500.dry
102
     0.008923
                          None
                500.dry
103
     0.008923
                         None
                100.dry
104
     0.008923
                         None
                100.dry
105
     0.008923
                         None
106
     0.008923
                100.dry
                         None
107
     0.008923
                100.dry
                         None
108
     0.008923
                100.dry
                         None
109
     0.008923
                100.dry
                         None
                100.dry
110
     0.008923
                         None
                100.dry
111
     0.008923
                         None
                100.dry
112
     0.008923
                         None
113
     0.008923
                100.dry
                         None
114
     0.008923
                100.dry
                         None
115
     0.008923
                100.dry
                         None
116
     0.008923
                100.dry
                         None
                100.dry None
117
     0.008923
                                                  energy spec_num
0
     [-0.0810641, -0.0721407, -0.0632173, -0.054293...
     [-0.0810641, -0.0721407, -0.0632173, -0.054293...
1
                                                                4
2
     [-0.0810641, -0.0721407, -0.0632173, -0.054293...]
                                                                4
     [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                                4
3
4
     [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                                4
5
     [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                                4
     [-0.0810641, -0.0721407, -0.0632173, -0.054293...
6
                                                                4
7
     [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                                4
     [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                                4
8
9
     [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                                4
10
     [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                               14
     [-0.0810641, -0.0721407, -0.0632173, -0.054293...
11
                                                                2
12
     [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                                2
13
     [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                                2
     [-0.0810641, -0.0721407, -0.0632173, -0.054293...
14
                                                                6
     [-0.0810641, -0.0721407, -0.0632173, -0.054293...
15
                                                               12
     [-0.0810641, -0.0721407, -0.0632173, -0.054293...
16
                                                               10
17
     [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                                6
     [-0.0810641, -0.0721407, -0.0632173, -0.054293...
18
                                                               14
19
     [-0.0810641, -0.0721407, -0.0632173, -0.054293...]
                                                               12
20
     [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                               15
     [-0.0810641, -0.0721407, -0.0632173, -0.054293...
21
                                                               14
22
     [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                               10
     [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                                9
23
     [-0.0810641, -0.0721407, -0.0632173, -0.054293...
24
                                                                1
25
     [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                                6
26
     [-0.0810641, -0.0721407, -0.0632173, -0.054293...]
                                                               12
27
     [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                               15
```

```
[-0.0810641, -0.0721407, -0.0632173, -0.054293...
29
                                                1
    [-0.0810641, -0.0721407, -0.0632173, -0.054293...
88
                                                1
89
    [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                1
    [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                4
90
91
    [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                5
92
    [-0.0810641, -0.0721407, -0.0632173, -0.054293...]
                                                3
    [-0.0810641, -0.0721407, -0.0632173, -0.054293...
93
                                                4
94
    [-0.0810641, -0.0721407, -0.0632173, -0.054293...]
                                                2
    [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                1
95
    [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                4
96
                                                2
    [-0.0810641, -0.0721407, -0.0632173, -0.054293...
97
    [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                3
98
    [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                5
99
    [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                5
100
101
    [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                1
    [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                2
102
    [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                3
103
    [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                2
104
105
    [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                3
    [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                4
106
107
    [-0.0810641, -0.0721407, -0.0632173, -0.054293...]
                                                1
    [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                3
108
    [-0.0810641, -0.0721407, -0.0632173, -0.054293...]
                                                1
109
                                                4
    [-0.0810641, -0.0721407, -0.0632173, -0.054293...
110
    [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                4
111
                                                2
112
    [-0.0810641, -0.0721407, -0.0632173, -0.054293...
    [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                5
113
    [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                5
114
    [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                5
115
    [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                3
116
117
    [-0.0810641, -0.0721407, -0.0632173, -0.054293...
                                                1
                                    intence
                                                     current
    0
                                           4.30.15.16
                                                        100
1
    4.30.15.16
                                                        100
2
    4.30.15.16
                                                        100
3
    4.30.15.16
                                                        100
4
    4.30.15.16
                                                        100
    5
                                           4.30.15.16
                                                        100
    6
                                           4.30.15.16
                                                        100
7
    4.30.15.16
                                                        100
8
    4.30.15.16
                                                        100
9
    4.30.15.16
                                                        100
10
    4.30.15.16
                                                        100
11
    4.30.15.16
                                                        100
12
    4.30.15.16
                                                        100
```

[-0.0810641, -0.0721407, -0.0632173, -0.054293...

```
13
 100
               4.30.15.16
 14
               4.30.15.16
                    100
15
 4.30.15.16
                    100
16
 4.30.15.16
                    100
               4.30.15.16
 17
                    100
 4.30.15.16
18
                    100
19
 4.30.15.16
                    100
 20
               4.30.15.16
                    100
 4.30.15.16
21
                    100
 22
               4.30.15.16
                    100
23
 4.30.15.16
                    100
24
 4.30.15.16
                    100
25
 4.30.15.16
                    100
26
 4.30.15.16
                    100
 27
               4.30.15.16
                    100
28
 4.30.15.16
                    100
29
 4.30.15.16
                    100
. .
                    . . .
 4.30.15.16
88
                    100
89
 4.30.15.16
                    100
90
 4.30.15.16
                    100
 91
               4.30.15.16
                    100
92
 4.30.15.16
                    100
93
 4.30.15.16
                    100
94
 4.30.15.16
                    100
95
 4.30.15.16
                    100
 4.30.15.16
96
                    100
97
 4.30.15.16
                    100
 98
               4.30.15.16
                    100
99
 4.30.15.16
                    100
100
 4.30.15.16
                    100
101
 4.30.15.16
                    100
102
 4.30.15.16
                    100
103
 4.30.15.16
                    100
 4.30.15.16
104
                    100
 105
               4.30.15.16
                    100
 106
               4.30.15.16
                    100
107
 4.30.15.16
                    100
               4.30.15.16
 108
                    100
 109
               4.30.15.16
                    100
110
 4.30.15.16
                    100
 4.30.15.16
111
                    100
112
 4.30.15.16
                    100
 4.30.15.16
113
                    100
 114
               4.30.15.16
                    100
115
               4.30.15.16
 100
               4.30.15.16
116
 100
117
 4.30.15.16
                    100
```

```
a0 ftype
                                        Gauss.sqe_Ta
                                                        Gauss.sqe_Zn
                                                                        Gauss.sqe_Fe
0
    -0.081064
                 npks
                                             0.000000
                                                        0.00000e+00
                                                                        12782.834144
                            . . .
1
                                                                        12718.985814
    -0.081064
                 npks
                                         8565.473145
                                                        2.413427e+04
                            . . .
2
    -0.081064
                 npks
                                         14826.830173
                                                        1.093657e+04
                                                                        12344.963302
                            . . .
3
    -0.081064
                 npks
                                             0.000000
                                                        0.000000e+00
                                                                        15003.922805
                            . . .
4
    -0.081064
                 npks
                                          9478.980972
                                                        3.186378e+04
                                                                        14336.032861
                            . . .
    -0.081064
5
                 npks
                                             0.00000
                                                        0.000000e+00
                                                                        14644.273729
                            . . .
6
    -0.081064
                 npks
                                         28468.057715
                                                        0.00000e+00
                                                                        15026.333239
                            . . .
7
    -0.081064
                 npks
                                         14330.464954
                                                        1.634471e+04
                                                                        12570.108187
                            . . .
8
                                         16508.984275
                                                        0.00000e+00
                                                                        11836.602490
    -0.081064
                 npks
                            . . .
9
    -0.081064
                 npks
                                         22907.390573
                                                        0.00000e+00
                                                                        13016.098882
                            . . .
10
    -0.081064
                 npks
                                         12168.767084
                                                        3.801942e+04
                                                                        18084.984698
                            . . .
11
    -0.081064
                 npks
                                        33177.564573
                                                        2.008907e+04
                                                                        21999.748065
                            . . .
12
    -0.081064
                 npks
                                         15650.501361
                                                        1.249096e+04
                                                                        18613.290457
                            . . .
13
    -0.081064
                 npks
                                         12785.212189
                                                        1.873494e+04
                                                                        22103.950412
                            . . .
14
    -0.081064
                                         10428.422579
                                                        1.722302e+04
                                                                        22684.081868
                 npks
15
                                             0.000000
    -0.081064
                 npks
                                                        1.389143e+04
                                                                        18922.414174
    -0.081064
                                                                        21301.380619
16
                 npks
                                             0.000000
                                                        0.000000e+00
                            . . .
17
    -0.081064
                                         13077.411572
                                                        1.512358e+04
                                                                        23565.440852
                 npks
                            . . .
18
    -0.081064
                 npks
                                        31868.528409
                                                        0.00000e+00
                                                                        21237.126948
                            . . .
19
    -0.081064
                 npks
                                         11428.295962
                                                        1.293526e+04
                                                                        18322.848533
                            . . .
20
    -0.081064
                 npks
                                         23661.917943
                                                        0.00000e+00
                                                                        22463.688186
                            . . .
21
    -0.081064
                 npks
                                         13378.774225
                                                        1.360371e+04
                                                                        20879.819810
                            . . .
22
    -0.081064
                                         31885.211856
                                                        1.199114e+04
                                                                        22285.444182
                 npks
23
    -0.081064
                                             0.00000
                                                        3.646227e+04
                                                                        22089.834582
                 npks
                            . . .
24
    -0.081064
                 npks
                                             0.000000
                                                        2.907814e+04
                                                                        22729.781269
                            . . .
25
    -0.081064
                 npks
                                         15339.243153
                                                        0.000000e+00
                                                                        24067.769152
                            . . .
26
    -0.081064
                 npks
                                         12983.195261
                                                        1.532537e+04
                                                                        21118.226804
                            . . .
27
    -0.081064
                                             0.00000
                                                        1.694376e+04
                                                                        23351.603819
                 npks
                            . . .
28
    -0.081064
                                         22394.064946
                                                        0.000000e+00
                                                                        21708.438724
                 npks
                            . . .
29
    -0.081064
                                         14902.591479
                                                        4.196798e+04
                                                                        21527.987140
                 npks
                  . . .
88
    -0.081064
                 npks
                                         17459.891209
                                                        1.912141e+04
                                                                        23114.454086
                            . . .
89
    -0.081064
                                         11383.449332
                                                        2.699227e+04
                                                                        23974.740146
                 npks
                            . . .
90
    -0.081064
                 npks
                                         27756.386481
                                                        0.00000e+00
                                                                        22207.420937
                            . . .
91
    -0.081064
                 npks
                                         18034.165889
                                                        3.272106e+04
                                                                        24703.529740
                            . . .
92
    -0.081064
                                         37729.405193
                                                        1.481763e+04
                                                                        23667.133694
                 npks
                            . . .
93
    -0.081064
                 npks
                                         9738.995484
                                                        1.032582e+06
                                                                        22913.562346
                            . . .
94
    -0.081064
                 npks
                                        39778.856377
                                                        2.125088e+04
                                                                        25753.931879
95
    -0.081064
                                         11521.894943
                 npks
                                                        1.460006e+04
                                                                        23345.839991
                            . . .
96
    -0.081064
                 npks
                                             0.000000
                                                        0.00000e+00
                                                                        24356.422465
                            . . .
97
    -0.081064
                                         11387.062720
                 npks
                                                        1.673381e+04
                                                                        26305.593400
                            . . .
98
    -0.081064
                 npks
                                         16603.882690
                                                        1.652611e+04
                                                                        22415.363137
                            . . .
99
    -0.081064
                                         13032.133215
                                                        1.508637e+04
                                                                        24848.413533
                 npks
                            . . .
100 -0.081064
                 npks
                                         18338.691350
                                                        5.217903e+04
                                                                        24642.968019
                            . . .
101 -0.081064
                 npks
                                         38781.272262
                                                        3.045623e+04
                                                                        23923.754104
102 -0.081064
                                         16171.824932
                                                        0.00000e+00
                                                                        25702.831983
                 npks
```

```
103 -0.081064
                npks
                                       11943.336090
                                                      3.155645e+04
                                                                     22764.527801
                           . . .
104 -0.081064
                npks
                                       18682.224660
                                                      1.779901e+04
                                                                     23806.263290
105 -0.081064
                npks
                                       14980.769173
                                                      2.635840e+04
                                                                     23248.282644
                           . . .
                                       16465.815743
                                                                     23531.119674
106 -0.081064
                npks
                                                      1.928858e+04
                           . . .
107 -0.081064
                npks
                                       14689.546031
                                                      1.731407e+04
                                                                     22712.884956
                           . . .
108 -0.081064
                npks
                                       16085.003695
                                                      0.00000e+00
                                                                     23080.131174
                           . . .
109 -0.081064
                                      128767.486528
                                                      1.221977e+04
                                                                     23394.874398
                npks
                           . . .
110 -0.081064
                npks
                                       18470.250894
                                                      1.415890e+04
                                                                     24344.874604
                           . . .
111 -0.081064
                npks
                                       13892.623295
                                                      0.00000e+00
                                                                     23354.774451
                           . . .
112 -0.081064
                npks
                                       13524.013718
                                                      3.322714e+04
                                                                     23046.616163
                           . . .
                                                      1.534304e+04
113 -0.081064
                npks
                                       15836.481220
                                                                     23613.920101
                           . . .
114 -0.081064
                npks
                                       13079.930366
                                                      0.00000e+00
                                                                     23206.197311
                           . . .
115 -0.081064
                npks
                                       16985.393297
                                                      2.972103e+04
                                                                     23318.744100
                           . . .
116 -0.081064
                npks
                                       28068.057352
                                                      4.031932e+04
                                                                     22945.535363
                           . . .
117 -0.081064
                npks
                                       23263.906373
                                                      0.00000e+00
                                                                     22988.399475
                           . . .
      Gauss.sqe_S
                                    Gauss.sqe_Mo
                                                                    Gauss.sqe_Ca
                      Gauss.sqe_K
                                                   Gauss.sqe_Sr
0
      9950.664247
                                    10943.905700
                                                    66125.880105
                                                                    81113.835877
                     86934.671678
1
      9686.542485
                     85417.271458
                                    10321.926453
                                                    65017.214651
                                                                    78708.054755
2
     10116.839276
                     89172.181946
                                    11689.414083
                                                    65654.177592
                                                                    79611.013286
3
     10903.042334
                     95391.063701
                                    11901.029557
                                                    74405.676539
                                                                    89784.619791
4
     10985.610374
                     93685.496143
                                    11950.637694
                                                   74258.461860
                                                                    92481.740876
5
     10102.485332
                     92416.008506
                                    11919.901894
                                                   72988.272331
                                                                    87443.420675
6
     10173.684196
                     91750.911548
                                    12281.715791
                                                   73525.258635
                                                                    90682.880967
7
                                                    66762.569120
                                                                    77191.878749
      9400.197705
                     81235.316095
                                    10928.679308
8
                                                    63394.499320
                                                                    72392.560589
      9343.140147
                     83119.955300
                                     9867.151628
9
     10533.779147
                     92094.717260
                                    10863.713674
                                                    65003.522088
                                                                    77511.668721
10
     31461.396864
                    167520.416366
                                    12268.002918
                                                   71901.664924
                                                                   149028.532972
11
     34043.224119
                    176152.435083
                                    13059.701119
                                                   77588.835766
                                                                   165933.128640
12
     29767.615193
                    152361.238746
                                    11635.315627
                                                    69148.083974
                                                                  147868.989290
13
     36287.148514
                    185771.446650
                                    13912.330811
                                                   81824.942625
                                                                  178381.992779
                                                   89757.119138
14
     42149.038277
                    211846.018133
                                    14253.788002
                                                                  173856.056045
15
     32948.656074
                    174243.178389
                                    13054.029077
                                                   72548.982421
                                                                  148822.370711
16
     36554.802547
                    186139.808068
                                    13164.250198
                                                   81530.158191
                                                                  171933.176260
17
                    190932.863067
     37233.406984
                                    12380.715522
                                                   88081.186779
                                                                  169881.641382
18
     35993.463377
                    188605.028636
                                    13101.342033
                                                   84474.819191
                                                                   170201.238709
19
     32426.596807
                    172952.845998
                                    11791.312437
                                                   73407.814517
                                                                   146543.517828
20
     36225.292796
                    189496.419296
                                    13015.032919
                                                   81639.403055
                                                                  180091.003226
21
     35112.262728
                    189916.864769
                                    13295.047631
                                                   85000.068670
                                                                  170866.444743
22
     36598.560215
                    182827.615530
                                    13487.845749
                                                   80990.776490
                                                                  176756.362560
23
     33705.061034
                    167491.996045
                                    12187.912707
                                                   80558.920058
                                                                  163348.523937
24
     36841.433201
                    190034.759929
                                    13696.745871
                                                   80858.622457
                                                                   178694.244484
25
     36214.761519
                    181693.257026
                                    12838.485307
                                                   82557.406208
                                                                  173538.425609
26
     36605.342817
                    194600.250214
                                    14468.045145
                                                   81895.421177
                                                                   169043.320980
27
     35996.140988
                    189018.863597
                                    13651.978890
                                                   82163.386277
                                                                  180632.103208
28
     37293.792638
                    192135.027228
                                    14005.577355
                                                   82974.359212
                                                                  173838.559143
29
     36481.904855
                    190348.936709
                                    13239.906774
                                                   78154.024442
                                                                  174508.646166
```

. .

. . .

```
88
     38019.188032
                    208089.944924
                                   13197.902741
                                                  81318.882008
                                                                 187463.126934
89
     38107.612934
                    206724.963254
                                    13671.291461
                                                  82092.146036
                                                                 188299.953087
90
     37046.184161
                    203762.844922
                                    12358.572023
                                                  81783.830002
                                                                 185465.968543
91
     37748.446305
                    204782.863964
                                   15564.227574
                                                  81170.486111
                                                                 187106.810147
92
     37977.777617
                    206006.639031
                                    12770.644638
                                                  80635.567466
                                                                 186516.033975
93
     37819.737694
                    205008.110607
                                    13343.425714
                                                  81480.105825
                                                                 188134.459819
94
     37720.562112
                    203724.862252
                                    14420.529242
                                                  80370.124129
                                                                 184957.368465
95
     37941.343840
                    206511.142515
                                    12932.277570
                                                  81273.897915
                                                                 187187.513234
96
     37804.533156
                    204851.042700
                                   12769.935197
                                                  81440.647735
                                                                 186977.526070
97
     37828.757982
                    205211.894349
                                    12919.082029
                                                  80998.038636
                                                                 186956.824940
98
     37963.731304
                                                  80612.285504
                    205931.659146
                                    13445.818063
                                                                 186840.968117
99
     38002.976908
                    206353.618367
                                    14031.853472
                                                  79757.968622
                                                                 188105.259246
100
     37664.123914
                    206996.811196
                                    13463.996242
                                                  81409.568351
                                                                 187864.300852
101
     38148.476564
                    205136.844090
                                    13240.784584
                                                  81245.416868
                                                                 187329.365544
102
     37575.536382
                    204145.179049
                                    12677.680280
                                                  81103.426050
                                                                 185035.479327
103
     37597.435082
                    207146.384864
                                   12461.215256
                                                  81614.548163
                                                                 187691.783398
104
     38029.926437
                    205609.665196
                                   14119.208047
                                                  80518.003477
                                                                 187605.398882
105
     37988.328513
                    208208.189481
                                    13692.673967
                                                  81473.763889
                                                                 188174.428294
106
     38400.279334
                    204411.657769
                                    15321.356067
                                                  79530.716888
                                                                 186806.840036
107
     37866.838929
                    204752.032768
                                    13838.623040
                                                                 186801.192919
                                                  81455.108594
108
     37774.881399
                    206457.652699
                                    13112.448431
                                                  81100.927070
                                                                 188089.808321
109
     37692.590263
                    206212.742318
                                    14270.444349
                                                  81235.710182
                                                                 187420.734195
110
     37810.516746
                    205109.144999
                                    14845.410236
                                                  81066.724362
                                                                 187987.248988
111
     38083.973255
                    205496.757359
                                    13206.441670
                                                  82456.036979
                                                                 186168.653676
                    206341.081239
                                    13157.511496
                                                  81137.200479
112
     38103.534095
                                                                 187754.208056
                    208221.736640
113
     38228.023308
                                    14758.706129
                                                  81751.208621
                                                                 188845.293994
     37855.033772
                    206520.970325
                                   14196.722526
                                                  80743.848554
                                                                 187761.248975
114
115
     38021.713412
                    206403.039312
                                    14949.336374
                                                  80489.275024
                                                                 187917.469303
116
     38259.548801
                    206990.920068
                                    14032.602396
                                                  82260.697460
                                                                 188132.942594
117
     38312.579918
                    205952.483265
                                    13967.984447
                                                  80848.987220
                                                                 187073.574118
```

0 4909.040539 27369.833763 1 5373.082037 28551.141260 2 32451.903800 4916.262922 3 5521.977542 30136.133442 4 5885.501915 27987.210135 5 5412.528251 28721.179215 6 5841.398563 25459.707706 7 5129.791153 28071.229452 8 4613.960749 29372.695713 9 4988.685533 34847.103425 10 7359.053110 63792.954162 11 8490.775946 65236.485942 12 6990.959429 53891.413004 13 8524.783568 66098.786523 14 9183.700383 83323.830322 15 8074.282433 65223.889884

Gauss.sqe_Cl

Gauss.sqe_Mn

```
16
     8576.031235 69937.854381
17
     8818.906675 71755.672493
18
     8782.120916 71812.627351
     8062.035331 64617.729393
19
20
     8677.966954 68895.354372
21
     8984.545843
                 70510.672079
22
     8869.550305
                 65020.915585
23
     8096.842579 58331.152657
24
     8674.952322 66198.994369
25
     9542.865092 63906.489970
26
                 71695.627587
     9210.756032
27
     8415.914766 68536.737534
28
     8747.378116
                 74651.808178
29
     8398.805304
                 70051.554516
. .
             . . .
     9284.223825
                 75042.867553
88
89
     9103.238329
                 74256.929009
90
     9211.902317
                 73404.028755
91
     8927.476876 73630.200495
92
     8824.445010
                 73361.674099
93
     9069.743840
                 73740.247513
94
     8532.309231
                 73523.156349
95
     9289.213555 74102.178823
96
     8883.147261 73661.838849
97
     8660.337759 74062.309050
98
     9001.481216 73718.790865
99
     8791.001603
                 74428.625300
100
    9668.147833
                74191.456735
101
     8741.891873
                 73876.645995
102
    9329.836744 73821.919521
103
    8644.913008 73487.349456
104
     8976.735222 71327.791787
105
    8806.215639 71653.016004
    9207.220147 70690.912488
106
     9308.656898
107
                71413.879473
108
     9094.662508 70988.166469
109
     9254.857684 71115.376340
    8718.233572 70545.709839
110
111
    8692.061631 71241.391159
112 9211.608709 71550.266011
113 9409.090441 71090.017776
    8876.563170 71423.194114
114
115
    8695.165823
                 71165.166964
116
    9157.021773 71182.171158
    9125.117111 70681.796694
117
```

[118 rows x 64 columns]