Lab4

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1 SKOLTECH, Experimental Data Processing

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```
In [1]: import numpy as np
        import scipy as sp
        from matplotlib import pyplot as plt
        from numpy.linalg import inv
        import matplotlib as mplb
        from matplotlib.font_manager import FontProperties
        %matplotlib inline
        from numpy.random import normal
        from mpl_toolkits.mplot3d import Axes3D
        mplb.rc('xtick', labelsize=5)
        mplb.rc('ytick', labelsize=5)
In [2]: def generate_trajectory(sigma_w_2, sigma_n_2, M, X_0):
            sigma_w = sigma_w_2 ** 0.5
            sigma_n = sigma_n_2 ** 0.5
            w = np.random.normal(0, sigma_w, M)
            n = np.random.normal(0, sigma_n, M)
            x = w.dot(np.tril(np.ones((M, M)))) + X_0 * np.ones(M)
            z = x + n
            return x, z
        def calculate_exp_w_n(z):
            v = z - np.roll(z,1)
            p = z - np.roll(z, 2)
            E_v = np.average((v ** 2)[1:])
            E_p = np.average((p ** 2)[2:])
            return E_p - E_v, E_v - E_p / 2
        def alpha_opt(sigma_w_2, sigma_n_2):
            chi = sigma_w_2 / sigma_n_2
            return (-chi + (chi**2 + 4 * chi) ** 0.5) / 2
        def M_opt(alpha):
```

```
return int(round((2 - alpha) / alpha))
def running_mean(z, M, mask=None):
    if mask is None:
       mask = np.ones(M) / M
   m = int(M/2)
    z ext = np.zeros(len(z) + 2 * m)
    z_{ext}[:m] = np.flip(z[:m], 0)
    z_{ext}[-m:] = np.flip(z[-m:], 0)
    z_{ext}[m:-m] = z
    return np.convolve(z_ext, mask, mode="same")[m:-m]
def exp_mean(z, alpha):
    exp_mean_z = z.copy()
    for i in range(1, len(z)):
        exp_mean_z[i] = exp_mean_z[i-1] + alpha * (z[i] - exp_mean_z[i-1])
    return exp_mean_z
def backward_exp_mean(x_f, alpha):
    x b = x f.copy()
    for i in range(len(x_b)-2,-1,-1):
        x_b[i] = x_b[i+1] + alpha*(x_f[i] - x_b[i+1])
   return x b
```

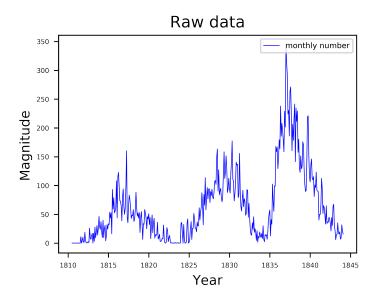
1.2 Part 1:

1.3 Comparison of the traditional 13 month running mean with the forward

1.4 backward exponential smoothing for approximation of 11 year sunspot cycle

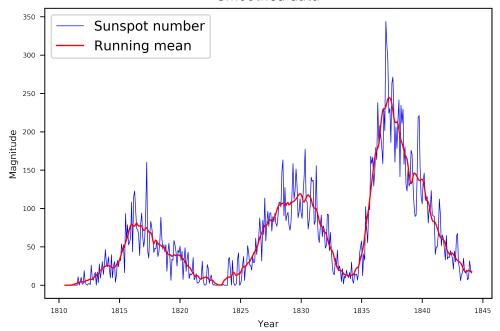
```
In [3]: year, month, y = np.loadtxt("data_group6.txt", unpack=True) # y - sunspot number
    x = year + month / 12 # x - months

In [4]: fig, ax = plt.subplots(1,1, figsize=(4,3), dpi = 600)
    ax.set_title('Raw data')
    ax.plot(x, y,'b', label = 'monthly number', linewidth=0.5)
    fontP = FontProperties()
    fontP.set_size('xx-small')
    ax.legend(prop = fontP, loc = 'upper right')
    ax.set_xlabel('Year')
    ax.set_ylabel('Magnitude');
```



1.4.1 Running mean

Smoothed data



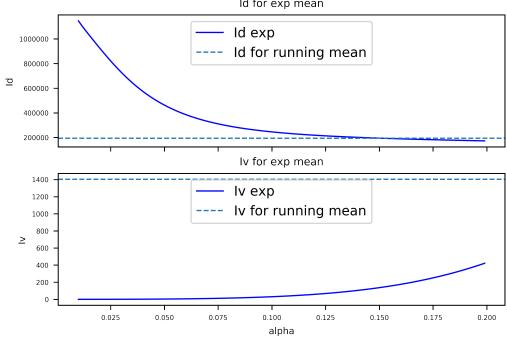
1.4.2 Exponential mean

```
In [7]: def Id(z, x):
            return np.sum((z-x)**2)
        def Iv(x):
            return np.sum((np.roll(x,-2) - 2 * np.roll(x,-1) + x)[:-2]**2)
In [8]: print("For running mean Id = %f, Iv = %f" % (Id(y, y_mean), Iv(y_mean)))
For running mean Id = 193951.331979, Iv = 1403.096250
In [9]: alphas = np.arange(0.01, 0.2, 0.001)
        Id_exp = []
       Iv_exp = []
        for alpha in alphas:
            y_exp_bf = backward_exp_mean(exp_mean(y, alpha), alpha)
            Id_exp.append(Id(y, y_exp_bf))
            Iv_exp.append(Iv(y_exp_bf))
In [10]: fig, ax = plt.subplots(2,1, figsize=(6,4), dpi = 600, sharex=True)
         ax[0].set_title('Id for exp mean', fontsize = 8)
         ax[0].plot(alphas, Id_exp,'b', label = 'Id exp', linewidth=1)
         ax[0].axhline(y=Id(y, y_mean), label = 'Id for running mean', linewidth=1, linestyle=
```

```
ax[0].set_ylabel('Id', fontsize = 7)
ax[0].legend(loc="upper center", fontsize=10);

ax[1].set_title('Iv for exp mean', fontsize = 8)
ax[1].plot(alphas, Iv_exp,'b', label = 'Iv exp', linewidth = 1)
ax[1].axhline(y=Iv(y_mean), label = 'Iv for running mean', linewidth=1, linestyle='da:
ax[1].set_xlabel('alpha', fontsize = 7)
ax[1].set_ylabel('Iv', fontsize = 7)
ax[1].legend(loc="upper center", fontsize=10);

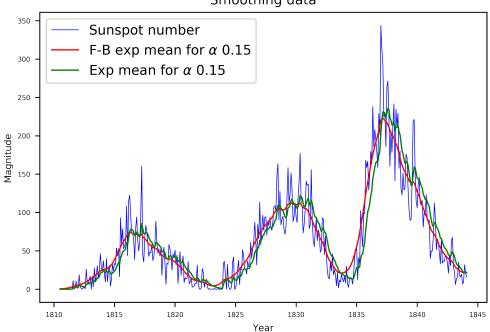
Id for exp mean
```



- 1.4.3 As it shown on the plot Iv coefficient for running mean is always more than same coefficient for forward-backward exponential mean.
- 1.4.4 And Id coefficient for running mean is smaller than same coefficient for forward-backward exponential mean when alpha is smaller than 0.15.
- 1.4.5 So if we take alpha > 0.15 we get both Id and Iv coefficient better for forward-backward exponential mean than running mean.

```
ax.plot(x, y,'b', label = r'Sunspot number', linewidth = 0.5 )
ax.plot(x, y_exp_bf,'r', label = r'F-B exp mean for $\alpha$ %.2f' % (alpha), linewid:
ax.plot(x, y_exp,'g', label = r'Exp mean for $\alpha$ %.2f' % (alpha), linewidth = 1
ax.set_xlabel('Year', fontsize=7)
ax.set_ylabel('Magnitude', fontsize=7)
ax.legend(loc = 'upper left', fontsize=10);
```

Smoothing data



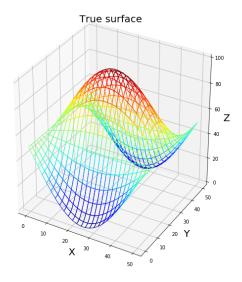
1.5 Part 2:

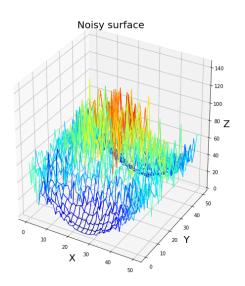
1.6 3d surface filtration using forward-backward smoothing

y, x = np.meshgrid(np.arange(N), np.arange(N))

```
In [12]: def fb_exp_mean_surface(z, alpha):
    z_fb_exp = z.copy()
    for i in range(z.shape[0]):
        z_fb_exp[i] = backward_exp_mean(exp_mean(z_fb_exp[i], alpha), alpha)
    z_fb_exp_mean_tr = z_fb_exp.transpose()
    for i in range(z.shape[0]):
        z_fb_exp_mean_tr[i] = backward_exp_mean(exp_mean(z_fb_exp_mean_tr[i], alpha),
        return z_fb_exp_mean_tr.transpose()
In [13]: z_true = np.loadtxt("true_surface.txt")
    z_noisy = np.loadtxt("noisy_surface.txt")
    N = z_true.shape[0]
```

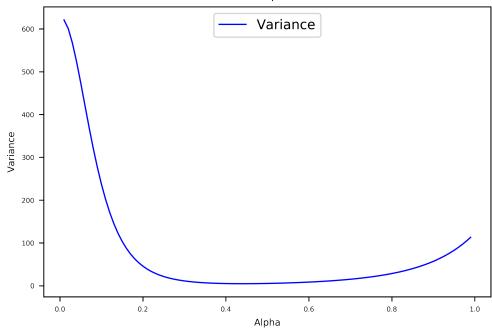
```
In [14]: def colors_for_graph(z):
             return mplb.cm.jet((z-z.min())/(z.max()-z.min()))
In [15]: fig = plt.figure(figsize=(20, 10))
         ax = fig.add_subplot(121, projection='3d')
         ax.set_title('True surface', fontsize = 20)
         surf = ax.plot_surface(x, y, z_true, facecolors = colors_for_graph(z_true), shade=Fale
         surf.set_facecolors((0, 0, 0, 0))
         ax.xaxis.set_tick_params(labelsize=10)
         ax.yaxis.set_tick_params(labelsize=10)
         ax.zaxis.set_tick_params(labelsize=10)
         ax.set_xlabel('X', fontsize=20)
         ax.set_ylabel('Y', fontsize=20)
         ax.set_zlabel('Z', fontsize=20)
         ax = fig.add_subplot(122, projection='3d')
         surf = ax.plot_surface(x, y, z_noisy, facecolors = colors_for_graph(z_noisy), shade=Facecolors
         surf.set_facecolor((0,0,0,0))
         ax.set_title('Noisy surface', fontsize = 20)
         ax.xaxis.set_tick_params(labelsize=10)
         ax.yaxis.set_tick_params(labelsize=10)
         ax.zaxis.set_tick_params(labelsize=10)
         ax.set_xlabel('X', fontsize=20)
         ax.set_ylabel('Y', fontsize=20)
         ax.set_zlabel('Z', fontsize=20);
```





In [16]: print("Variance for noisy data is %f" % (np.var(z_noisy - z_true)))

Variance from alpha for surface

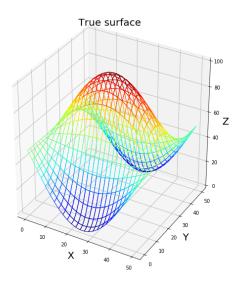


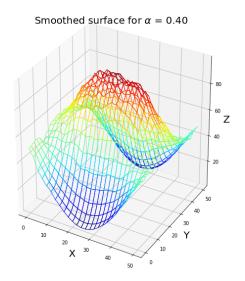
1.6.1 From graph it is seen that variance is small from 0.3 to 0.7

```
In [19]: alpha = 0.4
    fig = plt.figure(figsize=(20, 10))

ax = fig.add_subplot(121, projection='3d')
    ax.set_title('True surface', fontsize = 20)
    surf = ax.plot_surface(x, y, z_true, facecolors = colors_for_graph(z_true), shade=Falseurf.set_facecolors((0, 0, 0, 0)))
```

```
ax.xaxis.set_tick_params(labelsize=10)
ax.yaxis.set_tick_params(labelsize=10)
ax.zaxis.set_tick_params(labelsize=10)
ax.set_xlabel('X', fontsize=20)
ax.set_ylabel('Y', fontsize=20)
ax.set_zlabel('Z', fontsize=20)
ax = fig.add_subplot(122, projection='3d')
z_smooth = fb_exp_mean_surface(z_noisy, alpha)
surf = ax.plot_surface(x, y, z_smooth, facecolors = colors_for_graph(z_smooth), shade
surf.set_facecolor((0,0,0,0))
ax.set_title(r'Smoothed surface for $\alpha$ = %.2f' % (alpha), fontsize = 20)
ax.xaxis.set_tick_params(labelsize=10)
ax.yaxis.set_tick_params(labelsize=10)
ax.zaxis.set_tick_params(labelsize=10)
ax.set_xlabel('X', fontsize=20)
ax.set_ylabel('Y', fontsize=20)
ax.set_zlabel('Z', fontsize=20);
```





2 Today we learn forward-drawback exp mean for surface

In []: