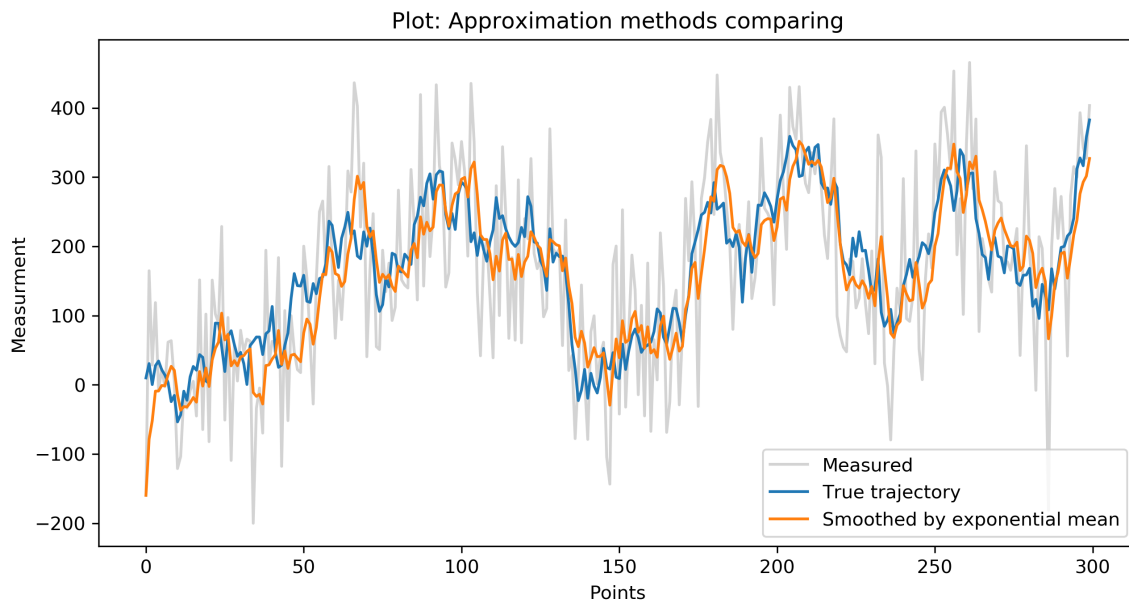


Assignment 3 Part I

In [9]:

```
#Assignment 2 Part 6
```

```
figure(num=None, figsize=(10, 5), dpi=300, facecolor='w', edgecolor='k')  
plt.title('Plot: Approximation methods comparing')  
plt.plot(z, label='Measured',c='lightgrey')  
plt.plot(x, label='True trajectory')  
plt.plot(sm_z, label='Smoothed by exponential mean')  
plt.xlabel('Points')  
plt.ylabel('Measurment')  
plt.legend()  
plt.show()
```

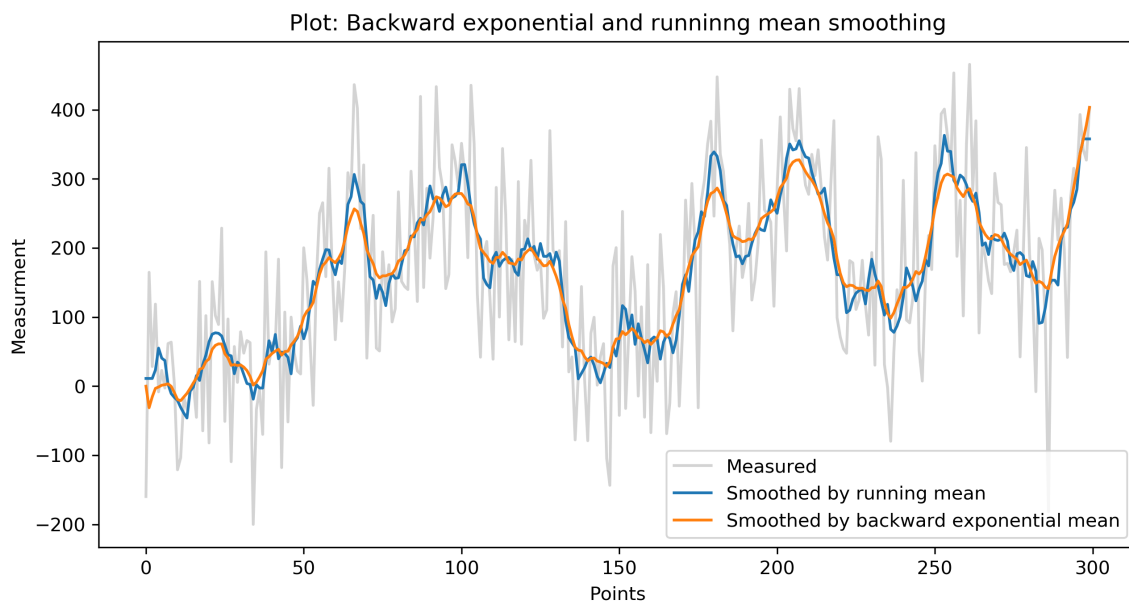


In [15]:

```
# Assignment 3
# Relationship between solar radio flux F10.7 and sunspot number
# Team 2:
#     Ekaterina Karmanova
#     Timur Chikichev
#     Yaroslav Okunev
#     Nikita Mikhailovskiy
#
# Skoltech, 04.10.2019

#backward smoothing
sm_z_back = np.zeros(c)
length = len(sm_z_back)
sm_z_back[length-1]=z[length-1]
for i in range (length-2,0,-1):
    sm_z_back[i] = sm_z_back[i+1] + a*(sm_z[i]-sm_z_back[i+1])

figure(num=None, figsize=(10, 5), dpi=300, facecolor='w', edgecolor='k')
plt.title('Plot: Backward exponential and running mean smoothing')
plt.plot(z, label='Measured', c='lightgrey')
plt.plot(sm_z_r, label='Smoothed by running mean')
plt.plot(sm_z_back, label='Smoothed by backward exponential mean')
plt.xlabel('Points')
plt.ylabel('Measurment')
plt.legend()
plt.show()
```



In [14]:

```
def devInd(z,x):
    Id=[0,0]
    for i in range(0,len(z)):
        Id[0]+=(z[i]-x[i])**2
    for i in range(0,len(z)-2):
        Id[1]+=(x[i+2]-2*x[i+1]+x[i])**2
    return Id
Idex=devInd(z,sm_z)
print('Indicator for forward exp. smoothing =',Idex)
Idbex=devInd(z,sm_z_back)
print('Indicator for backward exp. smoothing =',Idbex)
```

Indicator for forward exp. smoothing = [2126522.1308989194, 442778.4085260831]
Indicator for backward exp. smoothing = [2409214.250412533, 18484.320823410028]

In []:

```
1
2 import numpy as np
3
4 class Random_walking_model():
5     @staticmethod
6     def TrueTrajectory(initial, diffs):
7         array = [initial]
8         for di in diffs:
9             array.append(array[-1] + di)
10        return np.array(array)
11
12    def generate_signal(self, initial, sigma, n):
13        diffs = np.random.normal(0, sigma, n-1)
14        self.x = Random_walking_model.TrueTrajectory(initial, diffs)
15
16    def __init__(self, initial, sigma, n):
17        self.sigma = sigma
18        self.generate_signal(initial, sigma, n)
19
20 def rvm_alpha(sig):
21     '''linear regression coefficient(alpha)
22     for random valking model process'''
23     psi = sig[0]**2 / sig[1]**2
24
25     alpha = (-psi + np.sqrt(psi**2 + 4 * psi)) / 2
26     print('Optimal smoothing coefficient=', alpha)
27     return alpha
28
```

```

1
2 # Assignment 3
3 # Determining and removing drawbacks of exponential and running mean.
4 # Team 2:
5 #     Ekaterina Karmanova
6 #     Timur Chikichev
7 #     Iaroslav Okunevich
8 #     Nikita Mikhailovskiy
9 #
10 # Skoltech, 04.10.2019
11
12 import numpy as np
13
14 class Smooth():
15     def __init__(self, z):
16         self.z = z
17
18 class RunningMean(Smooth):
19     @staticmethod
20     def smooth(z, m):
21         x = np.zeros(len(z))
22         sumFirst = 0
23         sumLast = 0
24         step = int((m-1)/2)
25         for i in range(0, step):
26             sumFirst += z[i]
27             sumLast += z[len(z)-i-1]
28         for i in range(step, len(z)-step):
29             for j in range(i-step, i+step+1):
30                 x[i] += z[j]
31             x[i] /= m
32         for i in range(0, step):
33             x[i] = sumFirst/step
34             x[len(x)-i-1] = sumLast/step
35         return x
36
37     @staticmethod
38     def window_width(alpha):
39         return round((2-alpha)/alpha)
40
41     # def calc_window_width(self, alpha):
42     #     self.m = RunningMean.window_width(alpha)
43
44     def __init__(self, z, alpha):
45         super().__init__(z)
46         # Smooth.__init__(z)
47         self.m = RunningMean.window_width(alpha)
48         # return m
49
50     def run(self, m):
51         ret = RunningMean.smooth(self.z, m)
52         return ret
53
54 class FB_exp_smoothing(Smooth):
55     @staticmethod
56     def forward(alpfa, z):
57         array = np.zeros(len(z))
58         array[0] = 10
59         for i in range(1, len(z)):
60             array[i] = array[i - 1] + alpfa * (z[i] - array[i - 1])
61         return array

```

```
62
63 @staticmethod
64 def backward(alpha, fz):
65     array = np.zeros_like(fz)
66     array[len(fz)-1] = fz[len(fz)-1]
67     for i in range(len(fz)-2, -1, -1):
68         array[i] = array[i+1]+alpha*(fz[i]-array[i+1])
69     return array
70
71 def run_f(self, alpha):
72     res = FB_exp_smoothing.forward(alpha, self.z)
73     return res
74
75 def run_b(self, f, alpha):
76     res = FB_exp_smoothing.backward(alpha, f)
77     return res
78
79 def run_fb(self, alpha):
80     forward = FB_exp_smoothing.forward(alpha, self.z)
81     res = FB_exp_smoothing.backward(alpha, forward)
82     return res
83
```

```

1 # Assignment 3
2 # Determining and removing drawbacks of exponential and running mean.
3 # Team 2:
4 #     Ekaterina Karmanova
5 #     Timur Chikichev
6 #     Iaroslav Okunevich
7 #     Nikita Mikhailovskiy
8 #
9 # Skoltech, 04.10.2019
10
11 import os
12 try:
13     os.chdir(os.path.join(os.getcwd(), 'lab3'))
14     print(os.getcwd())
15 except:
16     pass
17
18 import numpy as np
19 import matplotlib.pyplot as plt
20 from matplotlib.pyplot import figure
21
22 import smoothing as sm
23 from walking_model import Random_walking_model, rvm_alpha
24
25 class Measurements():
26     def __init__(self, x, sigma):
27         n = len(x)
28         noise = np.random.normal(0, sigma, n)
29         self.z = x + noise
30
31     def DevVarInd(self, x):
32         #find deviation and variability indicators.
33         Id = 0
34         Iv = 0
35         z = self.z
36         for i in range(len(z)):
37             Id += (z[i]-x[i])**2
38         for i in range(0, len(z)-2):
39             Iv += (x[i+2] - 2*x[i+1] + x[i])**2
40         return Id, Iv
41
42
43 def draw_plots(plots=[], show = False):
44     for pi, plabel in plots:
45         plt.plot(pi, label=plabel)
46
47     plt.legend()
48     if show:
49         plt.show()
50
51 def new_plot(title, xl, yl, plots = [], show = False):
52     figure(num=None, figsize=(10, 5), dpi=100,
53           facecolor='w', edgecolor='k')
54     plt.title(title)
55     plt.ylabel(yl)
56     plt.xlabel(xl)
57
58     draw_plots(plots, show)
59
60 #Part II

```

```

61 def part2():
62     a = np.random.normal(0, np.sqrt(10), 300)
63     sig = np.sqrt(500)
64
65     def Trajectory(size, acc, t):
66         vel = Random_walking_model.TrueTrajectory(0, acc * t)
67         trajectory = Random_walking_model.TrueTrajectory(5, vel[:-1]*t + acc * t * t / 2)
68
69         new_plot('Plot result', 'Measurement', 'Points',
70                 [
71                     [acc, 'Acceleration'],
72                     [vel, 'Velocity'],
73                     [trajectory, 'Trajectory']
74                 ], show = False)
75         plt.savefig(fname='Trajectory.png')
76         return trajectory
77
78     traject = Trajectory(300, a, 0.1)
79     Known_vals = Measurements(traject, sig)
80     z = Known_vals.z
81
82     RM = sm.RunningMean(z, 1)
83
84     new_plot('Plot result',
85             'Measurements', 'Points', [
86                 [traject, 'Trajectory'],
87                 [z, 'Measure']
88             ], show = False)
89     plt.savefig(fname = 'running mean 1.png')
90
91     def eval_params(z, chvals, b, run):
92         krits = []
93         for ai in chvals:
94             r = run(ai)
95             ki = Known_vals.DevVarInd(r)
96             res = np.sum([ki[i]*b[i] for i in range(2)])
97             print('a: {} ki: {} res: {}'.format(
98                 ai, [ki[i]*b[i] for i in range(2)], res))
99             krits.append(res)
100         idmin = np.argmin(krits, 0)
101         ret = chvals[idmin - 1: idmin + 2]
102         return ret, krits[idmin]
103
104     def plot_f(z, chvals, run, aname):
105         new_plot('Plot result', 'Number', 'Count')
106         plt.plot(z, label='Measure', c='lightgrey')
107         for ai in chvals:
108             r = run(ai)
109             ki = Known_vals.DevVarInd(r)
110             res = np.sum([ki[i]*b[i] for i in range(2)])
111             plt.plot(r, label='{}, val: {}, res: {}'.format(aname, ai, res))
112         plt.legend()
113         # plt.show()
114
115     n = 10
116     b = [0.02, 0.08]
117     chvals = np.round(np.linspace(0, z.shape[0] * 0.4, n))[1:]
118     bestm, resm = eval_params(z, chvals, b, RM.run)
119     plot_f(z, bestm, RM.run, 'Running mean')
120     plt.savefig(fname='Running mean best.png')

```



```
121
122     Expsmoothing = sm.FB_exp_smoothing(z)
123
124     probe_alpha = np.linspace(0, 1, n + 2)[1:-1]
125     best_alpha, resalpha = eval_params(z, probe_alpha, b, Expsmoothing.run_fb)
126
127     print("bestm: {}, best_alpha: {}".format(bestm, best_alpha))
128     plot_f(z, best_alpha, Expsmoothing.run_fb, 'exp fb smoothing')
129
130     plt.savefig(fname='exp fb smoothing best.png')
131
132     def proper_choise(m_vals, alpha_vals):
133         m = m_vals
134         a = alpha_vals
135
136         res = 'exponential' if a < m else 'running mean'
137         print("mean:{}, exp:{}".format(m_vals, alpha_vals))
138         print(res)
139
140     proper_choise(resm, resalpha)
141
142 def main():
143     part2()
144
145 if __name__ == "__main__":
146     main()
147
148
149
```

Assignment 3 Part II - Second trajectory

In [1]:

```
# Assignment 3 Part 2 Trajectory 2
# Relationship between solar radio flux F10.7 and sunspot number
# Team 2:
#     Ekaterina Karmanova
#     Timur Chikichev
#     Yaroslav Okunev
#     Nikita Mikhailovskiy
#
# Skoltech, 04.10.2019
```

In [2]:

```
import numpy as np
import matplotlib.pyplot as plt
from matplotlib.pyplot import figure
```

In [3]:

```
#4 second trajectory
#initial acceleration
A1 = 1
#N. of points
C = 200
#Period counts
T = 32
#Variance
sw = 0.08
s2w = sw**2

#noise
w = np.random.normal(0,sw,C)
#Angle frequency
afreq = 2*np.pi/T

#determine acceleration
A = np.zeros(C)
A[0] = A1
for i in range (1,C):
    A[i] = A[i-1] + w[i]

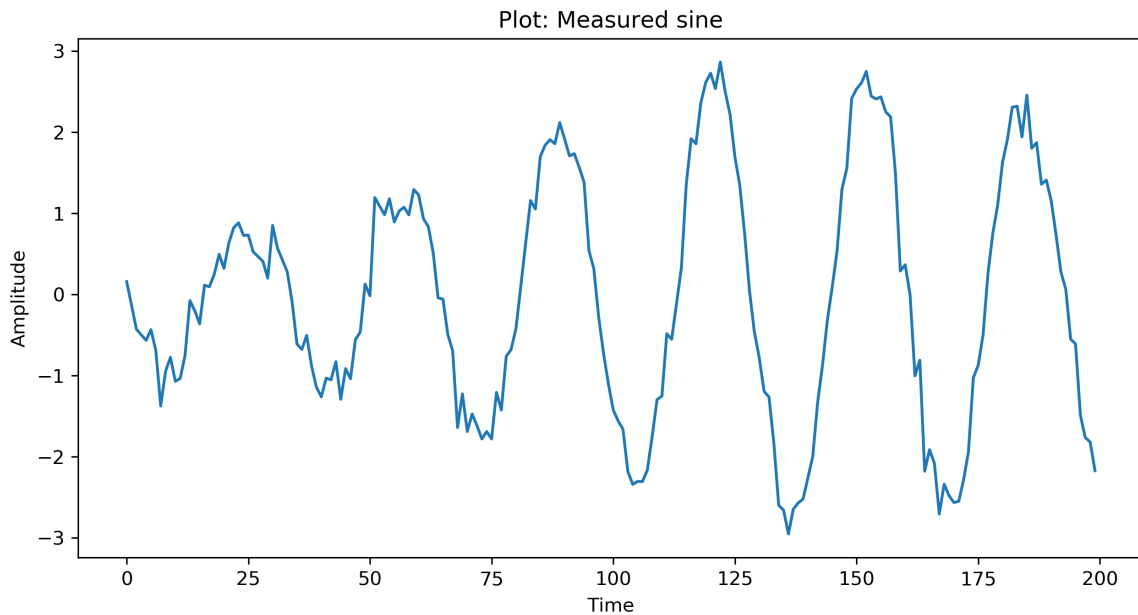
X = np.zeros(C)
for i in range (C):
    X[i] = A[i]*np.sin(afreq*i+3)
```

In [4]:

```
#5 Generating measurments
#Variance
s2n = 0.05
sn = np.sqrt(s2n)
Z = np.zeros(C)
n = np.random.normal(0,sn,C)

#determine measurment
for i in range (C):
    Z[i] = X[i] + n[i]

figure(num=None, figsize=(10, 5), dpi=300, facecolor='w', edgecolor='k')
plt.title('Plot: Measured sine')
plt.plot(Z)
plt.xlabel('Time')
plt.ylabel('Amplitude')
plt.show()
```



In [5]:

```
#6 Running mean
def rm(array, m):
    length = array.size
    smooth_arr = np.empty([length])
    diff = int((m-1)/2)

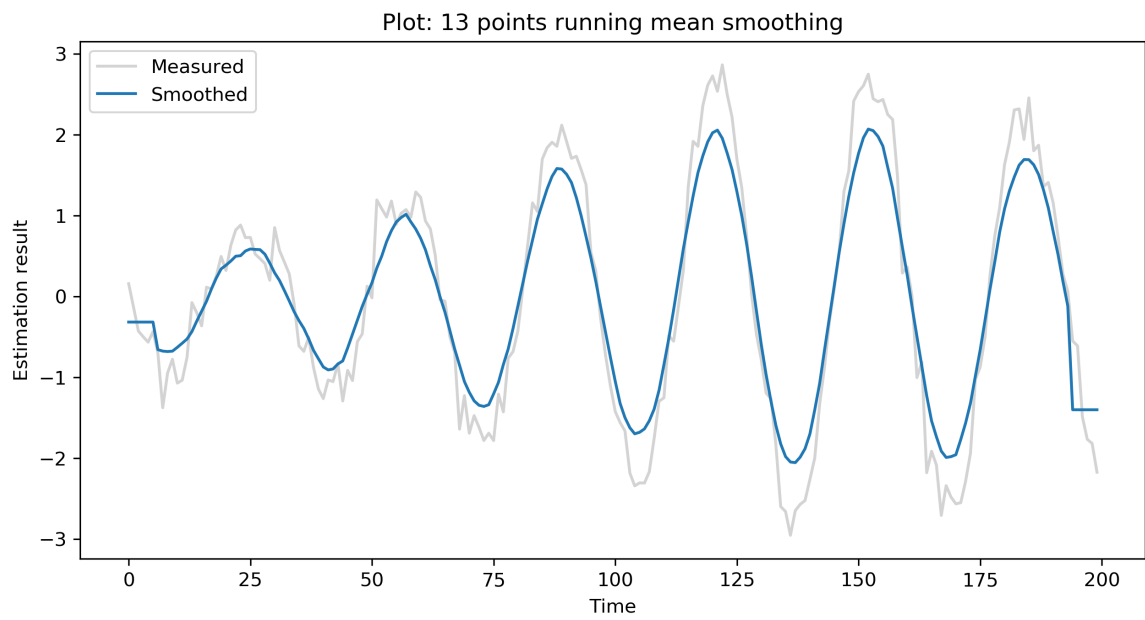
    #Pre-calculation: mean of the first six entries in dataset
    mean_first = 0
    for i in range (diff):
        mean_first += array[i]
    mean_first /= diff

    #Pre-calculation: mean of the last six entries in dataset
    mean_last = 0
    for i in range (diff):
        mean_last += array[length - i - 1]
    mean_last /= diff

    #Calculation for smoothing data in array including M period points
    for i in range (0,length):
        if i < diff:
            smooth_arr[i] = mean_first
        elif i > (length - diff-1):
            smooth_arr[i] = mean_last
        else:
            sum_it = 0
            for n in range (m):
                it = n - diff
                sum_it += 1/m*array[i-it]
            smooth_arr[i] = sum_it
    return smooth_arr

sm_z = rm(Z,13)

figure(num=None, figsize=(10, 5), dpi=300, facecolor='w', edgecolor='k')
plt.title('Plot: 13 points running mean smoothing')
plt.plot(Z, label='Measured', c='lightgrey')
plt.plot(sm_z, label='Smoothed')
plt.xlabel('Time')
plt.ylabel('Estimation result')
plt.legend()
plt.show()
```



In [6]:

```
#7 Period of oscillation  
M = 17
```

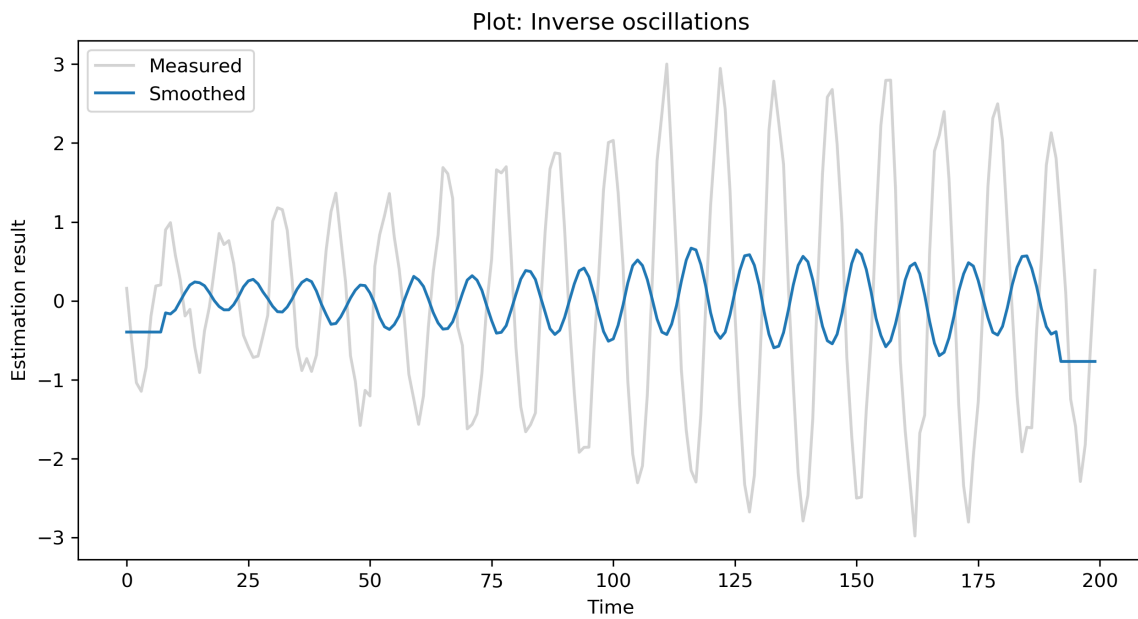
In [7]:

```
#a 1.5*M
afreq_a = 2*1.5*np.pi/M
#b M
afreq_b = 2*np.pi/M
#c < 0.5M
afreq_c = 2*0.4*np.pi/M

#a
Xa = np.zeros(C)
for i in range (C):
    Xa[i] = A[i]*np.sin(afreq_a*i+3)
#determine measurment
Za = np.zeros(C)
for i in range (C):
    Za[i] = Xa[i] + n[i]

sm_zs = rm(Za,M)

figure(num=None, figsize=(10, 5), dpi=300, facecolor='w', edgecolor='k')
plt.title('Plot: Inverse oscillations')
plt.plot(Za, label='Measured', c='lightgrey')
plt.plot(sm_zs, label='Smoothed')
plt.xlabel('Time')
plt.ylabel('Estimation result')
plt.legend()
plt.show()
```

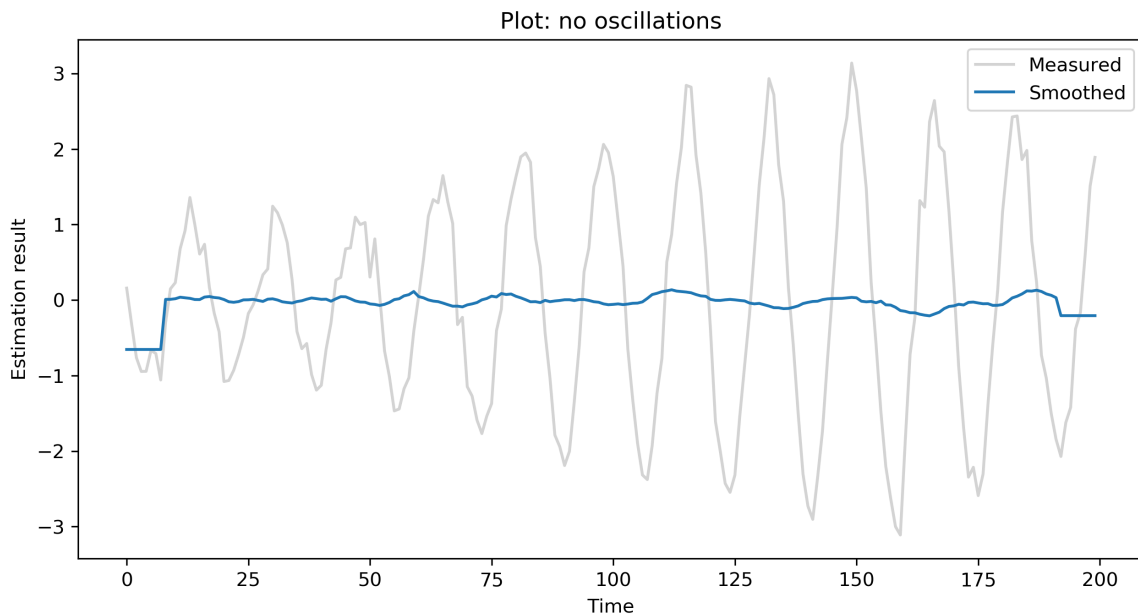


In [8]:

```
#b
Xb = np.zeros(C)
for i in range (C):
    Xb[i] = A[i]*np.sin(afreq_b*i+3)
#determine measurment
Zb = np.zeros(C)
for i in range (C):
    Zb[i] = Xb[i] + n[i]

sm_zb = rm(Zb,M)

figure(num=None, figsize=(10, 5), dpi=300, facecolor='w', edgecolor='k')
plt.title('Plot: no oscillations')
plt.plot(Zb, label='Measured', c='lightgrey')
plt.plot(sm_zb, label='Smoothed')
plt.xlabel('Time')
plt.ylabel('Estimation result')
plt.legend()
plt.show()
```

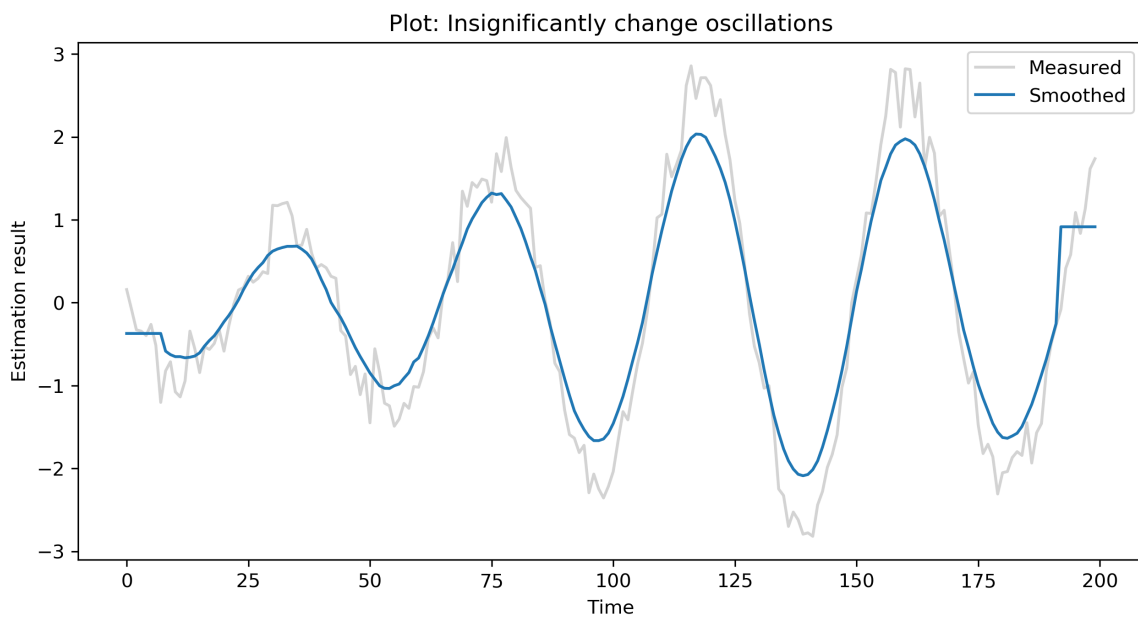


In [9]:

```
#c
Xc = np.zeros(C)
for i in range (C):
    Xc[i] = A[i]*np.sin(afreq_c*i+3)
#determine measurment
Zc = np.zeros(C)
for i in range (C):
    Zc[i] = Xc[i] + n[i]

sm_zc = rm(Zc,M)

figure(num=None, figsize=(10, 5), dpi=300, facecolor='w', edgecolor='k')
plt.title('Plot: Insignificantly change oscillations')
plt.plot(Zc, label='Measured', c='lightgrey')
plt.plot(sm_zc, label='Smoothed')
plt.xlabel('Time')
plt.ylabel('Estimation result')
plt.legend()
plt.show()
```



In []: