

Homework 2

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ABSTRACT

This homework has three parts.

1. Theoretical and programming questions related to data attributes and digital convolution.
2. Analyzing a digital elevation map by applying a running average filter.
3. Analyzing a digital elevation map by applying derivative filters.

PREREQUISITES

Completing the computational part of this homework assignment requires

- Madagascar software environment available from <http://www.ahay.org/>
- L^AT_EX environment with SEGT_EX available from <http://www.ahay.org/wiki/SEGT_EX>

The homework code is available from the Madagascar repository by running

```
svn co https://github.com/ahay/src/trunk/book/geo384h/hw2
```

DATA ATTRIBUTES AND CONVOLUTION

You can either write your answers to theoretical questions on paper or edit them in the file `hw2/paper.tex`. Please show all the mathematical derivations that you perform.

1. The varimax attribute is defined as

$$\phi[\mathbf{a}] = \frac{N \sum_{n=1}^N a_n^4}{\left(\sum_{n=1}^N a_n^2 \right)^2} \quad (1)$$

Suppose that the data vector \mathbf{a} contains random noise: the data values a_n are independent and identically distributed with a zero-mean Gaussian distribution: $E[a_n] = 0$, $E[a_n^2] = \sigma^2$, $E[a_n^4] = 3\sigma^4$. Find the mathematical expectation of $\phi[\mathbf{a}]$.

2. Consider the parabolic filter $F(Z)$ defines as

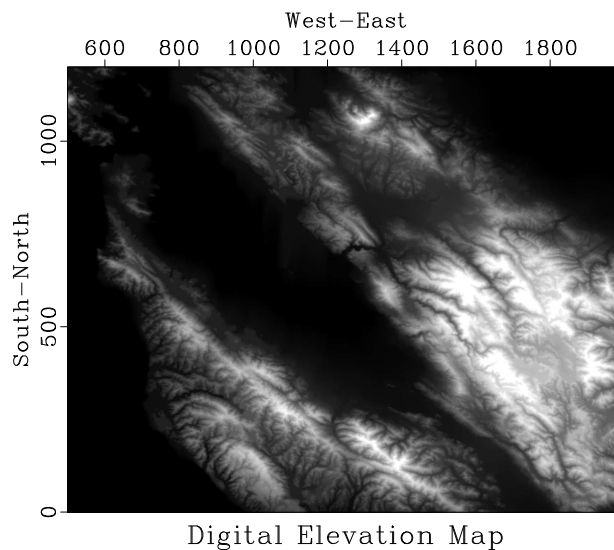
$$F(Z) = 1 + 4Z + 9Z^2 + \dots + N^2Z^{N-1} . \quad (2)$$

- (a) Show that this filter can be implemented using recursive filtering (polynomial division).
 - (b) What is the advantage of recursive filtering? Does it depend on N ?
3. Show that, using the helix transform and imposing helical boundary conditions, it is possible to compute a 2-D digital Fourier transform using 1-D FFT program. Assuming the input data is of size $N \times N$, would this approach have any computational advantages?

RUNNING MEDIAN AND RUNNING MEAN FILTERS

Figure 1: Digital elevation map of the San Francisco Bay Area.

[running/ bay](#)



We return the digital elevation map of the San Francisco Bay Area, shown in Figure ??.

In this exercise, we will separate the data into “signal” and “noise” by applying running mean and median filters. The result of applying a running median filter is shown in Figure 2. Running median effectively smooths the data by removing local outliers.

The algorithm is implemented in programs below.

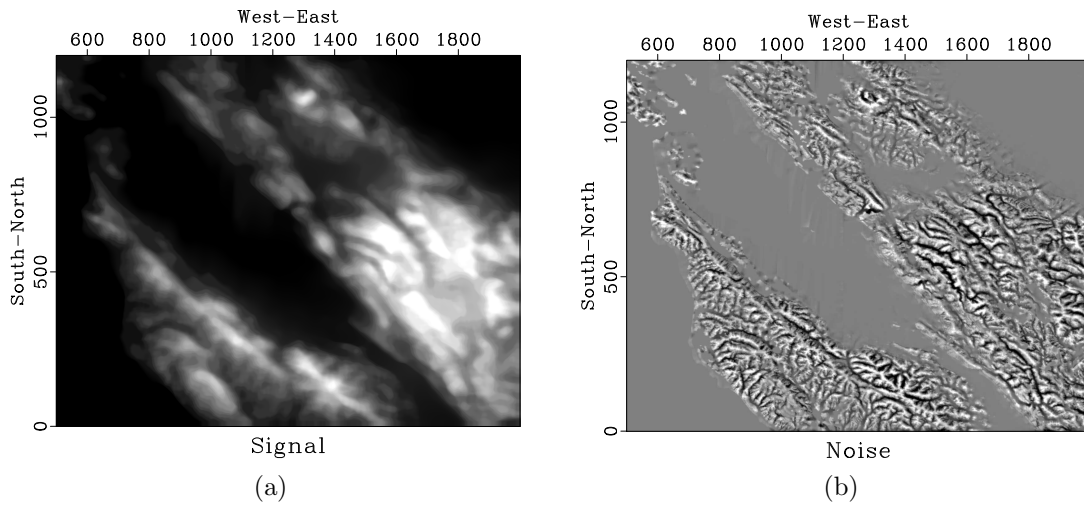


Figure 2: Data separated into signal (a) and noise (b) by applying a running median filter. [running/ave,res](#)

```

                                running/run.c
1  /* Apply running mean or median filter */
2
3  #include <rsf.h>
4
5  static float slow_median(int n, float* list)
6  /* find median by slow sorting, changes list */
7  {
8      int k, k2;
9      float item1, item2;
10
11     for (k=0; k < n; k++) {
12         item1 = list[k];
13
14         /* assume everything up to k is sorted */
15         for (k2=k; k2 > 0; k2--) {
16             item2 = list[k2-1];
17             if (item1 >= item2) break;
18             list[k2] = item2;
19         }
20         list[k2] = item1;
21     }
22
23     return list[n/2];
24 }
25

```

```

26 int main(int argc, char* argv [])
27 {
28     int w1, w2, nw, s1, s2, j1, j2, i1, i2, i3, n1, n2, n3;
29     char *how;
30     float **data, **signal, **win;
31     sf_file in, out;
32
33     sf_init (argc, argv);
34     in = sf_input("in");
35     out = sf_output("out");
36
37     /* get data dimensions */
38     if (!sf_histint(in, "n1", &n1)) sf_error("No n1=");
39     if (!sf_histint(in, "n2", &n2)) sf_error("No n2=");
40     n3 = sf_leftsize(in, 2);
41
42     /* input and output */
43     data = sf_floatalloc2(n1, n2);
44     signal = sf_floatalloc2(n1, n2);
45
46     if (!sf_getint("w1", &w1)) w1=5;
47     if (!sf_getint("w2", &w2)) w2=5;
48     /* sliding window width */
49
50     nw = w1*w2;
51     win = sf_floatalloc2(w1, w2);
52
53     how = sf_getstring("how");
54     /* what to compute
55      (fast median, slow median, mean) */
56     if (NULL == how) how="fast";
57
58     for (i3=0; i3 < n3; i3++) {
59
60         /* read data plane */
61         sf_floatread(data[0], n1*n2, in);
62
63         for (i2=0; i2 < n2; i2++) {
64             s2 = SF_MAX(0, SF_MIN(n2-w2, i2-w2/2-1));
65             for (i1=0; i1 < n1; i1++) {
66                 s1 = SF_MAX(0, SF_MIN(n1-w1, i1-w1/2-1));
67
68                 /* copy window */
69                 for (j2=0; j2 < w2; j2++) {
70                     for (j1=0; j1 < w1; j1++) {

```

```

71         win[j2][j1] = data[s2+j2][s1+j1];
72     }
73 }
74
75     switch (how[0]) {
76         case 'f': /* fast median */
77             signal[i2][i1] =
78                 sf_quantile(nw/2,nw,win[0]);
79             break;
80         case 's': /* slow median */
81             signal[i2][i1] =
82                 slow_median(nw,win[0]);
83             break;
84         case 'm': /* mean */
85             /* !!! ADD CODE !!! */
86             break;
87         default:
88             sf_error("Unknown method \"%s\"",how);
89             break;
90     }
91 }
92 }
93
94     /* write out */
95     sf_floatwrite(signal[0],n1*n2,out);
96 }
97
98     exit(0);
99 }

```

running/run.py

```

1  #!/usr/bin/env python
2
3  import sys
4  import numpy as np
5  import m8r
6
7  def slow_median(data):
8      'find median by slow sorting, changes data'
9
10     n = len(data)
11     for k in range(n):
12         item1 = data[k]
13

```

```

14     # assume everything up to k is sorted
15     for k2 in range(k, -1, -1):
16         item2 = data[k2-1]
17         if item1 >= item2:
18             break
19         data[k2] = item2
20
21     data[k2] = item1
22
23     return data[n/2]
24
25 # initialization
26 par = m8r.Par()
27 inp = m8r.Input()
28 out = m8r.Output()
29
30 # get data dimensions
31 n1 = inp.int('n1')
32 n2 = inp.int('n2')
33 n3 = inp.leftsize(2)
34
35 # input and output
36 data = np.zeros([n2, n1], 'f')
37 signal = np.zeros([n2, n1], 'f')
38
39 # sliding window
40 w1 = par.int('w1', 5)
41 w2 = par.int('w2', 5)
42
43 nw = w1*w2
44 win = np.zeros([w2, w1], 'f')
45
46 how = par.string('how', 'fast')
47 # what to compute
48 # (fast median, slow median, mean)
49
50 for i3 in range(n3):
51     # read data plane
52     inp.read(data)
53
54     for i2 in range(n2):
55         sys.stderr.write("\r%d of %d" % (i2, n2))
56         s2 = max(0, min(n2-w2, i2-w2/2-1))
57         for i1 in range(n1):
58             s1 = max(0, min(n1-w1, i1-w1/2-1))

```

```

59
60     # copy window
61     win = data[s2:s2+w2, s1:s1+w1]
62
63     if how[0] == 'f': # fast median
64         signal[i2, i1] = np.median(win)
65     elif how[0] == 's': # slow median
66         signal[i2, i1] = slow_median(win.flatten())
67     elif how[0] == 'm': # mean
68         # !!! ADD CODE !!!
69         pass
70     else:
71         sys.stderr.write(
72             "Unknown method \"%s\"\n" % how)
73         sys.exit(1)
74     sys.stderr.write("\n")
75     # write out
76     out.write(signal)
77
78 sys.exit(0)

```

1. Change directory to `hw2/running`.
2. Run

```
scons view
```

to reproduce the figures on your screen.

3. Modify the `run.c` program (alternatively, `run.py` script) and the `SConstruct` file to compute running mean instead of running median. Compare the results.
4. **EXTRA CREDIT** for improving the efficiency of the running median algorithm. Run

```
scons time.vpl
```

to display a figure that compares the efficiency of running median computations using the slow sorting from function `median` in program `run.c` (or `run.py`) and the fast median algorithm. Your goal is to make the algorithm even faster. You may consider parallelization, reusing previous windows, other fast sorting strategies, etc.

running/SConstruct

```

1 from rsf.proj import *
2
3 # Download data
4 Fetch('bay.h', 'bay')
5
6 # Convert format
7 Flow('bay', 'bay.h',
8      ', ,',
9      dd form=native |
10     window f2=500 n2=1500
11     ', ,')
12
13 # Display
14 def plot(title):
15     return ' , , '
16     grey allpos=y title="%s" yreverse=n
17     label1=South-North label2=West-East
18     screenratio=0.8
19     ' , , ' % title
20
21 Result('bay', plot('Digital Elevation Map'))
22
23 # Program for running average
24 run = Program('run.c')
25
26 # COMMENT ABOVE AND UNCOMMENT BELOW IF YOU WANT TO USE PYTHON
27 # run = Command('run.exe', 'run.py', 'cp $SOURCE $TARGET')
28 # AddPostAction(run, Chmod(run, 0o755))
29
30 w = 30
31
32 # !!! CHANGE BELOW !!!
33 Flow('ave', 'bay %s' % run[0],
34     './${SOURCES[1]} w1=%d w2=%d how=fast' % (w,w))
35 Result('ave', plot('Signal'))
36
37 # Difference
38 Flow('res', 'bay ave', 'add scale=1,-1 ${SOURCES[1]}')
39 Result('res', plot('Noise') + ' allpos=n')
40
41 #####
42
43 import sys

```



```

44
45 if sys.platform=='darwin':
46     gtime = WhereIs('gtime')
47     if not gtime:
48         print("For computing CPU time, install gtime.")
49 else:
50     gtime = WhereIs('gtime') or WhereIs('time')
51
52 # slow or fast
53 for case in ('fast', 'slow'):
54
55     ts = []
56     ws = []
57
58     time = 'time-' + case
59     wind = 'wind-' + case
60
61     # loop over window size
62     for w in range(3,16,2):
63         itime = '%s-%d' % (time,w)
64         ts.append(itime)
65
66         iwind = '%s-%d' % (wind,w)
67         ws.append(iwind)
68
69     # measure CPU time
70     Flow(iwind, None, 'spike n1=1 mag=%d' % (w*w))
71     Flow(itime, 'bay %s' % run[0],
72         ', , ,
73         ( (%s -f "%S %U"
74         ./${SOURCES[1]} < ${SOURCES[0]}
75         w1=%d w2=%d how=%s > /dev/null ) 2>&1 )
76         > time.out &&
77         (tail -1 time.out;
78         echo in=time0.asc n1=2 data-format=ascii_float)
79         > time0.asc &&
80         dd form=native < time0.asc | stack axis=1 norm=n
81         > $TARGET &&
82         /bin/rm time0.asc time.out
83         ' , , ' % (gtime,w,w,case), stdin=0, stdout=-1)
84
85     Flow(time, ts, 'cat axis=1 ${SOURCES[1:%d]}' % len(ts))
86     Flow(wind, ws, 'cat axis=1 ${SOURCES[1:%d]}' % len(ws))
87
88     # complex numbers for plotting

```

```

89     Flow('c'+time,[wind,time],
90         ',')
91     cat axis=2 ${SOURCES[1]} |
92     transp |
93     dd type=complex
94     ',')
95
96 # Display CPU time
97 Plot('time','ctime-fast ctime-slow',
98     ',')
99     cat axis=1 ${SOURCES[1]} | transp |
100    graph dash=0,1 wanttitle=n
101    label2="CPU Time" unit2=s
102    label1="Window Size" unit1=
103    ',',view=1)
104
105 End()

```

DERIVATIVE FILTERS

In this part of the assignment, we will use a digital elevation map of the Mount St. Helens area, shown in Figure 3.



Digital Elevation Map

Figure 3: Digital elevation map of Mount St. Helens area. [X](#)

[helix/ data](#)

Figure 4 shows a directional derivative, a digital approximation to

$$\cos \alpha \frac{\partial}{\partial x_1} + \sin \alpha \frac{\partial}{\partial x_2}, \quad (3)$$

applied to the data. A directional derivative highlights the structure of the mountain as if illuminating it with a light source.

Figure 4: Directional derivative of elevation. [helix/ der](#)

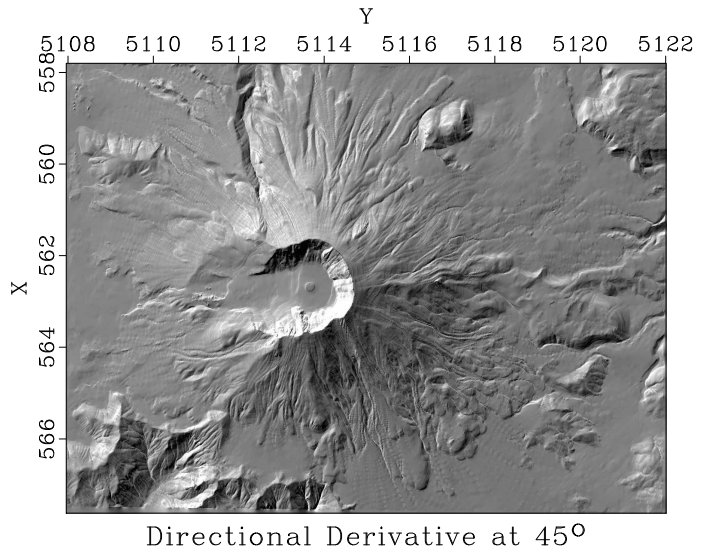


Figure 5 shows an application of *helical derivative*, a filter designed by spectral factorization of the Laplacian filter

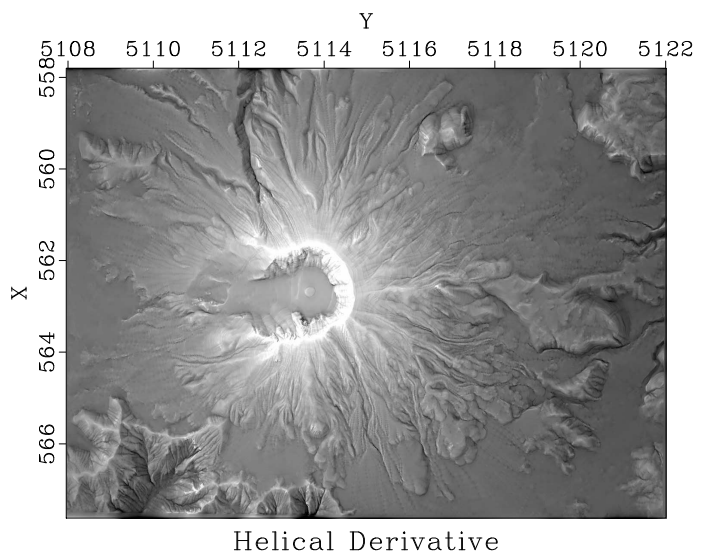
$$L(Z_1, Z_2) = 4 - Z_1 - 1/Z_1 - Z_2 - 1/Z_2 . \quad (4)$$

To invert the Laplacian filter, we put on a helix, where it takes the form

$$L_H(Z) = 4 - Z - Z^{-1} - Z^{N_1} - Z^{-N_1} , \quad (5)$$

and factor it into two minimum-phase parts $L_H(Z) = D(Z) D(1/Z)$ using the Wilson-Burg algorithm. The helical derivative $D(Z)$ enhances the image but is not confined to a preferential direction.

Figure 5: Helix derivative of elevation. [helix/ helder](#)



1. Change directory to hw2/helix.

2. Run

```
scons view
```

to reproduce the figures on your screen.

3. Edit the `SConstruct` file. Find the parameter that corresponds to α in equation (3) and try to modify it until you create the most interesting image. After changing the parameter, you can view the result by running

```
scons der.view
```

4. **EXTRA CREDIT** for suggesting and implementing a method for finding optimal α automatically.
5. A more accurate version of the Laplacian filter is

$$\hat{L}_2(Z_1, Z_2) = 20 - 4Z_1 - 4Z_1^{-1} - 4Z_2 - 4Z_2^{-1} - Z_1Z_2 - Z_1Z_2^{-1} - Z_2Z_1^{-1} - Z_1^{-1}Z_2^{-1}. \quad (6)$$

Modify the `SConstruct` file to use filter (6) instead of (4).

helix/SConstruct

```

1 from rsf.proj import *
2 import math
3
4 # Download data
5 txt = 'st-helens_after.txt'
6 Fetch(txt, 'data',
7       server='https://raw.githubusercontent.com',
8       top='agile-geoscience/notebooks/master')
9 Flow('data.asc', txt, '/usr/bin/tail -n +6')
10
11 # Convert to RSF format
12 Flow('data', 'data.asc',
13     '''
14     echo in=$SOURCE data.format=ascii_float
15     label=Elevation unit=m
16     n1=979 o1=557.805 d1=0.010030675 label1=X
17     n2=1400 o2=5107.965 d2=0.010035740 label2=Y |
18     dd form=native |
19     clip2 lower=0 | lapfill grad=y niter=200
20     '''
21
22 Result('data', 'grey title="Digital Elevation Map" allpos=y')
```

```

23
24 # Vertical and horizontal derivatives
25 Flow('der1','data','igrad')
26 Flow('der2','data','transp | igrad | transp')
27
28 ders = []
29 for alpha in range(0,360,15):
30     der = 'der%d' % alpha
31
32     # Directional derivative
33     Flow(der,'der1 der2',
34         '','
35         add ${SOURCES[1]} scale=%g,%g
36         ' ' % (math.cos(alpha*math.pi/180),
37               math.sin(alpha*math.pi/180)))
38     ders.append(der)
39
40 Flow('ders',ders,
41     '','
42     cat axis=3 ${SOURCES[1:%d]} |
43     put o3=0 d3=15
44     ' ' % len(ders))
45 Plot('ders','grey gainpanel=all wanttitle=n',view=1)
46
47 # !!! MODIFY BELOW !!!
48 alpha=45
49
50 Result('der','der%d' % alpha,
51     '','
52     grey title="Directional Derivative at %g\^o\_-"
53     ' ' % alpha)
54
55 # Laplacian filter on a helix (!!! MODIFY !!!)
56
57 Flow('slag0.asc',None,
58     ' 'echo 1 1000 n1=2 n=1000,1000
59     data_format=ascii_int in=$TARGET
60     ' ')
61 Flow('slag','slag0.asc','dd form=native')
62
63 Flow('ss0.asc','slag',
64     ' 'echo -1 -1 a0=2 n1=2
65     lag=$SOURCE in=$TARGET data_format=ascii_float ' ')
66 Flow('ss','ss0.asc','dd form=native')
67

```

```

68 # Wilson-Burg factorization
69
70 na=50 # filter length
71
72 lags = list(range(1,na+1)) + list(range(1002-na,1002))
73
74 Flow('alag0.asc',None,
75     '''echo %s n=1000,1000 n1=%d in=$TARGET
76     data_format=ascii_int
77     ''' % (' '.join([str(x) for x in lags]),2*na))
78 Flow('alag','alag0.asc','dd form=native')
79
80 Flow('hflt hlag','ss alag',
81     'wilson lagin=${SOURCES[1]} lagout=${TARGETS[1]}')
82
83 # Helical derivative
84
85 Flow('helder','data hflt hlag','helicon filt=${SOURCES[1]}')
86 Result('helder','grey mean=y title="Helical Derivative"')
87
88 def plotfilt(title):
89     return '''
90     window n1=11 n2=11 f1=50 f2=50 |
91     grey wantaxis=n title="%s" pclip=100
92     crowd=0.85 screenratio=1
93     ''' % title
94
95 # Laplacian impulse response
96 Flow('spike',None,'spike n1=111 n2=111 k1=56 k2=56')
97 Flow('imp0','spike ss','helicon filt=${SOURCES[1]} adj=0')
98 Flow('imp1','spike ss','helicon filt=${SOURCES[1]} adj=1')
99 Flow('imp','imp0 imp1','add ${SOURCES[1]}')
100 Plot('imp',plotfilt('(a) Laplacian'))
101
102 # Test factorization
103 Flow('fac0','imp hflt',
104     'helicon filt=${SOURCES[1]} adj=0 div=1')
105 Flow('fac1','imp hflt',
106     'helicon filt=${SOURCES[1]} adj=1 div=1')
107 Plot('fac0',plotfilt('(b) Laplacian/Factor'))
108 Plot('fac1',plotfilt('(c) Laplacian/Factor\''))
109 Flow('fac','fac0 hflt',
110     'helicon filt=${SOURCES[1]} adj=1 div=1')
111 Plot('fac',plotfilt('(d) Laplacian/Factor/Factor\''))
112

```

```
113 Result('laplace', 'imp fac0 fac1 fac', 'TwoRows',  
114         vppen='gridsize=5,5 xsize=11 ysize=11')  
115  
116 End()
```

COMPLETING THE ASSIGNMENT

1. Change directory to hw2.
2. Edit the file `paper.tex` in your favorite editor and change the first line to have your name instead of Hoare's.

3. Run

```
sftour scons lock
```

to update all figures.

4. Run

```
sftour scons -c
```

to remove intermediate files.

5. Run

```
scons pdf
```

to create the final document.

6. Submit your result (file `paper.pdf`.)