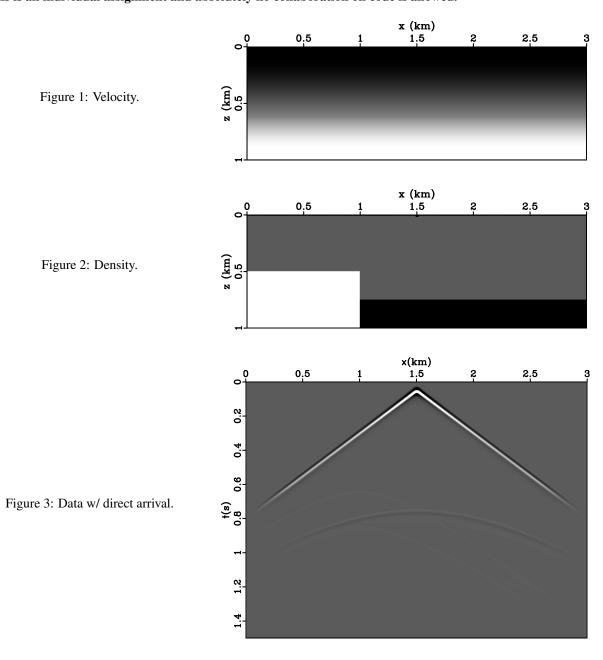
In this homework, you will use a finite-differences modeling code, similar to the one you wrote in the preceding homework, to implement basic reverse time migration. I do not expect you to be concerned with the efficiency of your implementation at this time. This implementation of reverse-time migration does not require that you write any new C code. You will use pre-existing Madagascar programs, but you will modify the SConstruct file to combine those programs.

This is an individual assignment and absolutely no collaboration on code is allowed.



EXERCISE

Using the finite-differences modeling function <code>awefd</code>, construct an image of the subsurface. This function takes the following parameters:

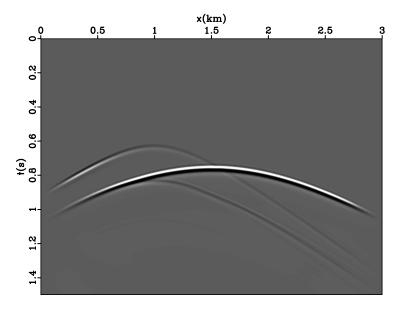


Figure 4: Data w/o direct arrival.

awefd(odat,owfl,idat,velo,dens,sou,rec,custom,par)

• odat: output data d(x,t)

• owfl: output wavefield u(z, x, t)

• idat: input data (wavelet)

• velo: velocity model v(z,x)

• dens: density model $\rho(z,x)$

• sou: source coordinates

• rec: receiver coordinates

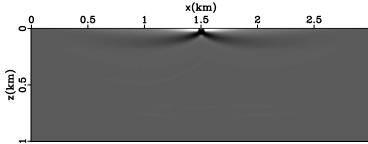
• custom: custom parameters

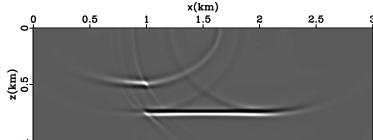
• par: parameter dictionary

Design an imaging procedure following the generic scheme developed in class. Your task is to identify Madagascar programs necessary to implement reverse-time migration in two different ways and generate the appropriate Flows in the SConstruct. Explain in detail how your imaging procedures work.

1. Use your imaging procedure to generate images based on recorded data in Figures 3 and 4. For this exercise, use the constant density rb.rsf for imaging. Include those two images in this document. Are the images different from each-other? How? Why?

Figure 5: Image migrated with the recorded data in Figure 3 and the constant density prb.rsf.



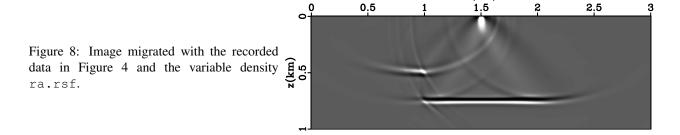


x(km)

Figure 6: Image migrated with the recorded data in Figure 4 and the constant density $\mathfrak{g}_{\mathbf{N}}$ of $\mathfrak{g}_{\mathbf{N}}$

2. Use your imaging procedure to generate images based on recorded data in Figures 3 and 4. For this exercise, use the variable density ra.rsf for imaging. Include those two images in this document. Are the images different from each-other? How? Why? How do your images compare with the ones from the preceding exercise?

Figure 7: Image migrated with the recorded data in Figure 3 and the variable density ra.rsf.



WRAP-UP

After you are satisfied that your document looks ok, print it from the PDF viewer and bring it to class.

SCONSTRUCT

```
# GPGN 658 - reverse-time migration
2 from rsf.proj import *
3 import fdm
              _____
4 # -----
5 par = dict(
6
    nt=1500, ot=0, dt=0.001, lt='t', ut='s',
7
     nx=601, ox=0, dx=0.005, lx='x', ux='km',
     nz=201, oz=0, dz=0.005, lz='z', uz='km',
8
9
     kt=50, nb=100, jsnap=50, jdata=1, frq=35
  )
10
11 fdm.param(par)
12
13 par['xk']=50
14 par['xl']=par['nx']-50
15
16 par['xsou']=par['ox']+par['nx']/2*par['dx']
17 par['zsou']=par['oz']
18
19 # -----
20 # wavelet
21 fdm.wavelet('wav_',par['frq'],par)
22 Flow( 'wav', 'wav_', 'transp')
23 Result('wav','window n2=500 | ' + fdm.waveplot('',par))
25 # -----
26 # sources coordinates
27 fdm.point('ss',par['xsou'],par['zsou'],par)
28 Plot('ss', fdm.ssplot('', par))
29
30 # receivers coordinates
31 fdm.horizontal('rr',0,par)
32 Plot('rr', fdm.rrplot('', par))
33
34 # -----
              _____
35 # velocity
36 Flow('vo', None,
   ,,,
37
38
     math output="2.0+0.25*x1"
39
     n1=%(nz)d o1=%(oz)g d1=%(dz)g
40
     n2=% (nx) d o2=% (ox) q d2=% (dx) q
      ''' % par)
41
42
43 Plot( 'vo', fdm.cgrey('allpos=y bias=2.0 pclip=100',par))
44 Result('vo',['vo','ss','rr'],'Overlay')
45
46 # -----
47 # density
48 Flow('ra', None,
      ,,,
49
50
      spike nsp=2 mag=+0.5, -0.5
51
      n1=%(nz)d o1=%(oz)q d1=%(dz)q k1=101,151 l1=%(nz)d, %(nz)d
52
     n2=% (nx) d o2=% (ox) g d2=% (dx) g k2=1,201 12=200,% (nx) d
53
     add add=2
54
     ''' % par)
55 Plot( 'ra', fdm.cgrey('allpos=y bias=1.5 pclip=100',par))
56 Result ('ra', ['ra', 'ss', 'rr'], 'Overlay')
57
58 Flow('rb','ra','math output=1')
59
60 # -----
```

```
61 # edge taper
62 Flow('taper', None,
63
64
         spike nsp=1 mag=1
65
         n1=%(nx)d d1=%(dx)g o1=%(ox)g k1=%(xk)d l1=%(xl)d
66
         n2=%(nt)d d2=%(dt)g o2=%(ot)g |
67
         smooth rect1=50
         ''' % par)
68
69
    Result('taper','transp |'+fdm.dgrey('pclip=99',par))
70
71
72 # finite-differences modeling
73 fdm.awefd('dd','ww','wav','vo','ra','ss','rr','jsnap=1 fsrf=n',par)
74 fdm.awefd('do','wo','wav','vo','rb','ss','rr','jsnap=1 fsrf=n',par)
75
76 Result('ww','window j3=%(jsnap)d |'%par + fdm.wgrey('pclip=99.9',par))
77 Result('wo','window j3=%(jsnap)d |'%par + fdm.wgrey('pclip=99.9',par))
78
79 # data w/ direct arrivals
80 Flow( 'dr0','dd taper',
           'add mode=p ${SOURCES[1]}')
81
82
83
   # data w/o direct arrivals
84 Flow( 'drl','dd do taper',
85
           'math r=\{SOURCES[0]\}\ d=\{SOURCES[1]\}\ t=\{SOURCES[2]\}\ output="(r-d)*t"')
86
87 for j in range (2):
88
        dtag="%d"%j
89
        Result('dr'+dtag,'transp | ' + fdm.dgrey('pclip=99.9',par))
91
92
    # Reverse-time migration
93
    imags = ['imag0','imag1']
    odats = ['odat0','odat1']
94
    tdats = ['tdat0','tdat1']
95
    twfls = ['twfl0','twfl1']
96
97
    rwfls = ['rwfl0','rwfl1']
98
    velo, sou, rec, custom='vo','rr','rr','jsnap=1 fsrf=n'
99
    dens, swfls, idats = ['ra','rb'], ['ww', 'wo'], ['dr0', 'dr1']
100
    xcor2dPar = 'xcor2d uu=${SOURCES[1]} axis=3 verb=y nbuf=100'
101
    for den in dens:
102
     for i in range(2):
103
        Flow(odats[i]+den,idats[i],'reverse which=2 opt=i verb=y')
104
        fdm.awefd(tdats[i]+den,twfls[i]+den,odats[i]+den,velo,den,sou,rec,custom,par)
105
        Flow(rwfls[i]+den,twfls[i]+den,'reverse which=4 opt=i verb=y')
106
        # conventional (cross-correlation zero-lag) imaging condition
        Flow(imags[i]+den,rwfls[i]+den+' '+swfls[i],xcor2dPar)
107
108
        Result(imags[i]+den,'window j3=%(jsnap)d |'%par + fdm.wgrey('pclip=99.9',par))
109
110 End()
```