

Appendix

```
%load data
load rowscan.mat
load diag1scan.mat
load diag2scan.mat
load std.mat

%row a, extract travel time
t_row = rowscan(1:16,5);
%diagonal_c, extract travel time
t_diag1 = diag1scan(1:31,5);
%diagonal_d, extract travel time
t_diag2 = diag2scan(1:31,5);

t = [t_row; t_diag1; t_diag2];

%model
G1 = zeros(16,256);

%row scan
%parameter indices increase by 17 each column and by 1 each row
for i = 1:16
    for j = 0:15
        if i+j*16 <= 256
            G1(i,i+j*16) = 1;
        else
            break
        end
    end
end
G1;

%diagonal scans
%upper part of the SW to NE
G2 = zeros(16,256);
for i = 1:16
    for j = 0:i-1
        if i+j*15 <= 256
            G2(i,i+j*15) = sqrt(2);
        else
            break
        end
    end
end

%lower part of the SW to NE
G3 = zeros(15,256);
for i = 1:15
    for j = 0:i-1
        if 256-(i-1)*16+j*15 <= 256
```

```

        G3(i, 256-(i-1)*16+j*15) = sqrt(2);
    else
        break
    end
end
end
end

```

```

G3_new = flipud(G3);

```

```

%lowerpart of the NW to SE

```

```

G4 = zeros(16,256);

```

```

for i = 1:16

```

```

    for j = 0:i-1

```

```

        if (16-(i-1)+j*17 <= 256)

```

```

            G4(i,16-(i-1)+j*17) = sqrt(2);

```

```

        else

```

```

            break

```

```

        end

```

```

    end

```

```

end

```

```

%upper part of the NW to SE

```

```

G5 = zeros(15,256);

```

```

for i = 1:15

```

```

    for j = 0:i-1

```

```

        if (241-(i-1)*16+j*17 <=256)

```

```

            G5(i,241-(i-1)*16+j*17) = sqrt(2);

```

```

        else

```

```

            break

```

```

        end

```

```

    end

```

```

end

```

```

G5_new = flipud(G5);

```

```

G = [G1; G2; G3_new; G4; G5_new];

```

```

[m,n] = size(G); % size of G

```

```

find(G(78,:)==sqrt(2)); % checking the model matrix

```

4.(a) Note the rank of your G matrix that relates the data and model.

```

%rank of G

```

```

p = rank(G)

```

```

% %Using SVD

```

```

[U,S,V] = svd(G);

```

```

Gdagger = V(:,1:p)*inv(S(1:p,1:p))*U(:,1:p)';

```

```
mdagger = Gdagger*t;
```

4.b) i. State and discuss significance of the elements and dimensions of the data and model null spaces.

```
% data null space
data_null_space = U(:,p+1:m)
%dimension of the data null spaces
dim_data_null_space = m-p
% model null space
V(:,p+1:n)
%dimension of the model null spaces
dim_model_null_space = n-p
```

Plot and interpret at least one element of each space,

```
figure(1)
plot(U(:,p+1),'.','MarkerSize',20)
xlabel('i'); ylabel('Data null space')
title('Data null space')
figure(2)
clf
colormap('gray')
imagesc(reshape(V(:,p+1),16,16))
set(colorbar,'FontSize',18);
```

```
xlabel('i'); ylabel('j')
title('Model null space')
```

ii. Contour or otherwise display a nonzero model that fits the trivial data set exactly

```
m_o = 6*V(:,p+1) + 5*V(:,p+2);
```

```
mm = reshape(m_o,16,16);
```

```
figure(3)
clf
colormap('gray')
imagesc(mm)
set(colorbar,'FontSize',18);
set(gca,'xtick',[1,2,3,4,5,6,7,8,9,10,11,12,13,14,15,16]);
set(gca,'ytick',[1,2,3,4,5,6,7,8,9,10,11,12,13,14,15,16]);
xlabel('j')
ylabel('i')
title('Non-zero model that fits the trivial dataset exactly')
```

iii. Show the model resolution by contouring or otherwise displaying the 256 diagonal elements of the model resolution matrix, reshaped into an appropriate 16 by 16 grid. Note if there are any model parameters that have perfect resolution.

```
% model resolution matrix
Vp=V(:,1:p);
Rm=Vp*Vp';
```

```
Rm = reshape(diag(Rm),16,16);
```

```
figure(4)
clf
colormap('gray')
```

```
imagesc(Rm)
set(colorbar,'FontSize',18);
xlabel('j')
ylabel('i')
title('Diagonal Model resolution matrix R_{m}')
The model parameters corresponding to the four edges (m_1, m_16, m_241, and m_256) have perfect resolutions.
```

(c) Produce a 16 by 16 element contour or other plot of your slowness perturbation model, displaying the maximum and minimum slowness perturbations in the title of the plot. Interpret any internal structures geometrically and in terms of seismic velocity (in m/s).

```
md = reshape(mdagger,16,16);
```

```
max_md = max(max(md));
min_md = min(min(md));
```

```
figure(5)
clf
colormap('gray')
imagesc(md)
set(colorbar,'FontSize',18);
```

```
title(['Slowness Perturbation model (slow_{max}=',num2str(max_md),', slow_{min}=',num2str(min_md),')'])
```

In addition, produce a plot of the data fit and discuss it.

```
data = G*mdagger;
```

```
figure(6)
plot(data, '*'); hold on
plot(t, 'o');
xlabel('i'); ylabel('Data fit')
title('Data fit')
legend('data fit', 'true data', Location='best')
```

(d) Describe how one could use solutions to $Gm = d = 0$ to demonstrate that very rough models exist that will fit any data set just as well as a generalized inverse model.

```
mw = md + 50*reshape(V(:,p+1),16,16);
```

```
max_ms = max(max(mw));
min_ms = min(min(mw));
```

```

figure(9)
clf
colormap('gray')
imagesc(mw)
set(colorbar,'FontSize',18);

title(['Wild Slowness Pertubation model (slow_{max}=',num2str(max_ms),', slow_{min}
=',num2str(min_ms),')'])

```

(e) Plot and interpret the correlation matrix.

```

Si = diag(S);
C = 0;
for i=1:p
    C = C + V(:,i)*V(:,i)'/Si(i)^2;
end
C = (std^2)*C;

```

```

Corr = corrcoef(C);

```

```

figure(7)
clf
colormap('gray')

imagesc(Corr)
set(colorbar,'FontSize',18);
xlabel('j')
ylabel('i')
title('Correlation Matrix')

```

(f) Quantify and discuss stability of slowness perturbation estimates.

```

CondGd = Si(1)/Si(p)

```

```

CondG = Si(1)/Si(78)

```