

4tile example

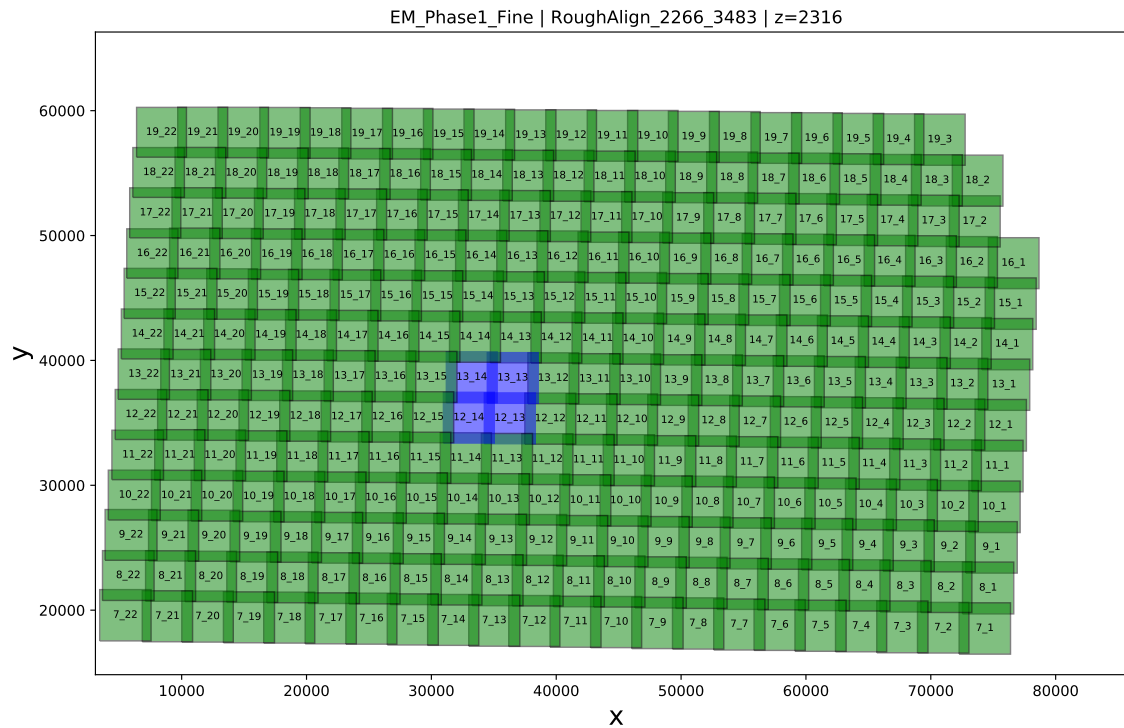
small example

- understand matrix construction, code and linear algebra

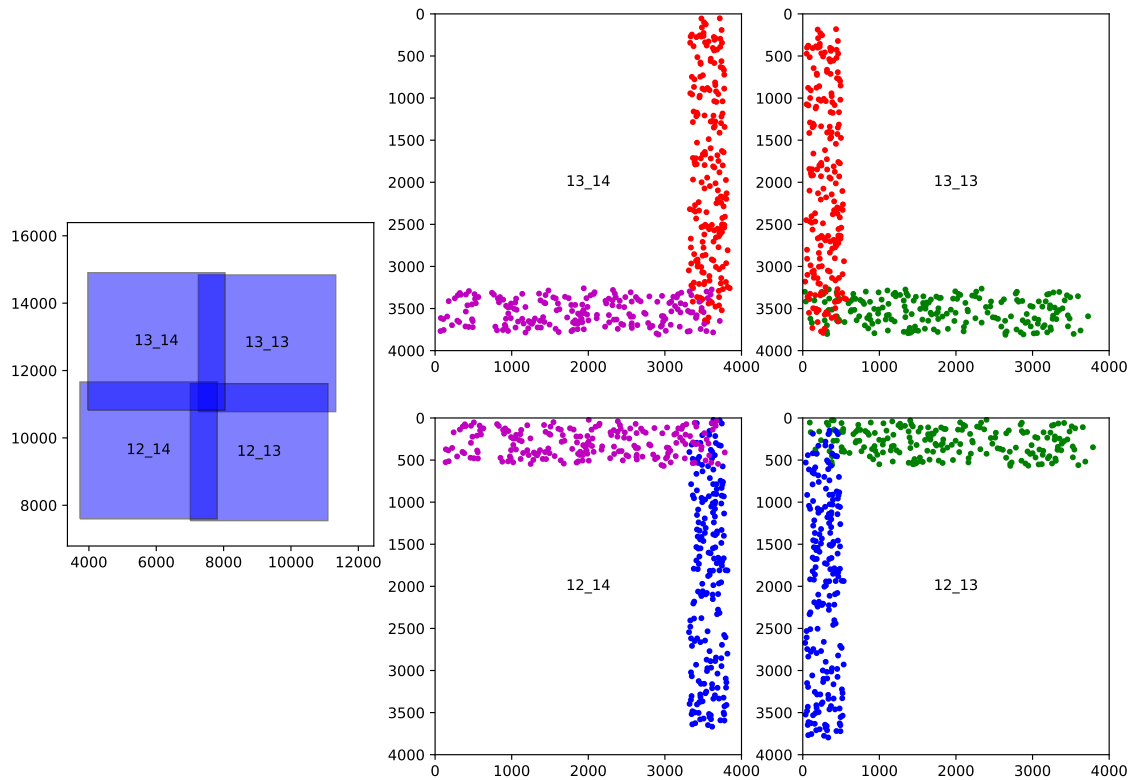
Dan Kapner

10/2/2017

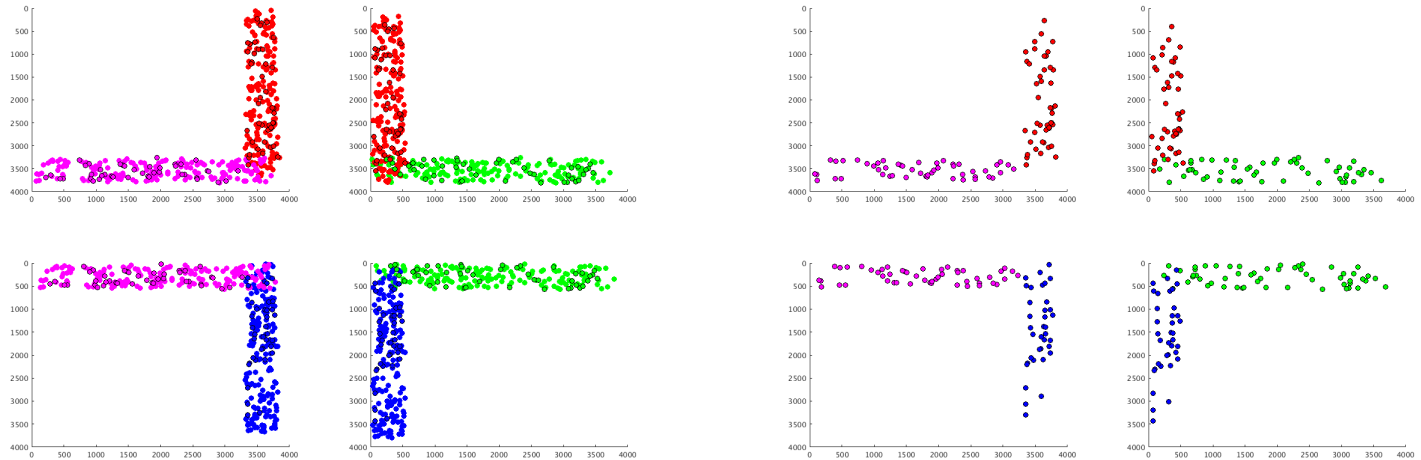
Source of Data



4 tiles and point matches



Filtering



MATLAB filters down the pointmatch collection from 200 points per tile pair to around 50
There is some random aspect to the filtering.

tile 1	tile 2
$(^1x_1, ^1y_1)$	$(^2x_1, ^2y_1)$
$(^1x_2, ^1y_2)$	$(^2x_2, ^2y_2)$
$(^1x_3, ^1y_3)$	$(^2x_3, ^2y_3)$
$(^1x_4, ^1y_4)$	$(^2x_4, ^2y_4)$
...	...

Naming and algebra

affine transformation

$$^1u_1 = a_1 ^1x_1 + b_1 ^1y_1 + c_1 \quad (1)$$

$$^1v_1 = d_1 ^1x_1 + e_1 ^1y_1 + f_1 \quad (2)$$

requirement for alignment

$$^1u_1 = ^2u_1 \quad (3)$$

$${}^1v_1 = {}^2v_1 \quad (4)$$

explicitly

$$a_1 {}^1x_1 + b_1 {}^1y_1 + c_1 = a_2 {}^2x_1 + b_2 {}^2y_1 + c_2 \quad (5)$$

$$d_1 {}^1x_1 + e_1 {}^1y_1 + f_1 = d_2 {}^2x_1 + e_2 {}^2y_1 + f_2 \quad (6)$$

and in matrix form, this is:

$$\begin{bmatrix} -{}^1x_1 & -{}^1y_1 & -1 & {}^2x_1 & {}^2y_1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & -{}^1x_1 & -{}^1y_1 & -1 & {}^2x_1 & {}^2y_1 & 1 \end{bmatrix} \times \begin{bmatrix} a_1 \\ b_1 \\ c_1 \\ a_2 \\ b_2 \\ c_2 \\ d_1 \\ e_1 \\ f_1 \\ d_2 \\ e_2 \\ f_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

and, now scaling to more than 1 point match pair:

$$\begin{bmatrix}
 -^1x_1 & -^1y_1 & -1 & ^2x_1 & ^2y_1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\
 -^1x_2 & -^1y_2 & -1 & ^2x_2 & ^2y_2 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\
 -^1x_3 & -^1y_3 & -1 & ^2x_3 & ^2y_3 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\
 -^1x_4 & -^1y_4 & -1 & ^2x_4 & ^2y_4 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\
 & & & \dots & & & & & & & & \\
 0 & 0 & 0 & 0 & 0 & 0 & -^1x_1 & -^1y_1 & -1 & ^2x_1 & ^2y_1 & 1 \\
 0 & 0 & 0 & 0 & 0 & 0 & -^1x_2 & -^1y_2 & -1 & ^2x_2 & ^2y_2 & 1 \\
 0 & 0 & 0 & 0 & 0 & 0 & -^1x_3 & -^1y_3 & -1 & ^2x_3 & ^2y_3 & 1 \\
 0 & 0 & 0 & 0 & 0 & 0 & -^1x_4 & -^1y_4 & -1 & ^2x_4 & ^2y_4 & 1 \\
 & & & & & & \dots & & & & &
 \end{bmatrix} \times \begin{bmatrix} a_1 \\ b_1 \\ c_1 \\ a_2 \\ b_2 \\ c_2 \\ d_1 \\ e_1 \\ f_1 \\ d_2 \\ e_2 \\ f_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ \dots \\ 0 \\ 0 \\ 0 \\ 0 \\ \dots \end{bmatrix}$$

we can rearrange the columns to keep all the tile coordinates in one

place:

$$\begin{bmatrix}
 -^1x_1 & -^1y_1 & -1 & 0 & 0 & 0 & ^2x_1 & ^2y_1 & 1 & 0 & 0 & 0 \\
 -^1x_2 & -^1y_2 & -1 & 0 & 0 & 0 & ^2x_2 & ^2y_2 & 1 & 0 & 0 & 0 \\
 -^1x_3 & -^1y_3 & -1 & 0 & 0 & 0 & ^2x_3 & ^2y_3 & 1 & 0 & 0 & 0 \\
 -^1x_4 & -^1y_4 & -1 & 0 & 0 & 0 & ^2x_4 & ^2y_4 & 1 & 0 & 0 & 0 \\
 & & & \dots & & & & & & & & \\
 0 & 0 & 0 & -^1x_1 & -^1y_1 & -1 & 0 & 0 & 0 & ^2x_1 & ^2y_1 & 1 \\
 0 & 0 & 0 & -^1x_2 & -^1y_2 & -1 & 0 & 0 & 0 & ^2x_2 & ^2y_2 & 1 \\
 0 & 0 & 0 & -^1x_3 & -^1y_3 & -1 & 0 & 0 & 0 & ^2x_3 & ^2y_3 & 1 \\
 0 & 0 & 0 & -^1x_4 & -^1y_4 & -1 & 0 & 0 & 0 & ^2x_4 & ^2y_4 & 1 \\
 & & & & & & \dots & & & & &
 \end{bmatrix} \times \begin{bmatrix} a_1 \\ b_1 \\ c_1 \\ d_1 \\ e_1 \\ f_1 \\ a_2 \\ b_2 \\ c_2 \\ d_2 \\ e_2 \\ f_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ \dots \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

this is starting to look like Khaled's presentation.

change all the signs (absorb into the a,b,c,d parameters)

$$[^{1,2}P \quad -^{1,2}Q] \times \begin{bmatrix} T_1 \\ T_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

where, for example

$$[{}^{1,2}P] = \begin{bmatrix} -{}^1x_1 & -{}^1y_1 & -1 & 0 & 0 & 0 \\ -{}^1x_2 & -{}^1y_2 & -1 & 0 & 0 & 0 \\ -{}^1x_3 & -{}^1y_3 & -1 & 0 & 0 & 0 \\ -{}^1x_4 & -{}^1y_4 & -1 & 0 & 0 & 0 \\ \dots & & & & & \\ 0 & 0 & 0 & -{}^1x_1 & -{}^1y_1 & -1 \\ 0 & 0 & 0 & -{}^1x_2 & -{}^1y_2 & -1 \\ 0 & 0 & 0 & -{}^1x_3 & -{}^1y_3 & -1 \\ 0 & 0 & 0 & -{}^1x_4 & -{}^1y_4 & -1 \\ & & & \dots & & \end{bmatrix}$$

and

$$T_1 = \begin{bmatrix} a_1 \\ b_1 \\ c_1 \\ d_1 \\ e_1 \\ f_1 \end{bmatrix}$$

and where the following is a column-wise concatenation

$$\begin{bmatrix} {}^{1,2}P & - {}^{1,2}Q \end{bmatrix}$$

and where the following is a row-wise concatenation

$$\begin{bmatrix} T_1 \\ T_2 \end{bmatrix}$$

this equation, again, is for 2 tiles:

$$\begin{bmatrix} {}^{1,2}P & - {}^{1,2}Q \end{bmatrix} \times \begin{bmatrix} T_1 \\ T_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

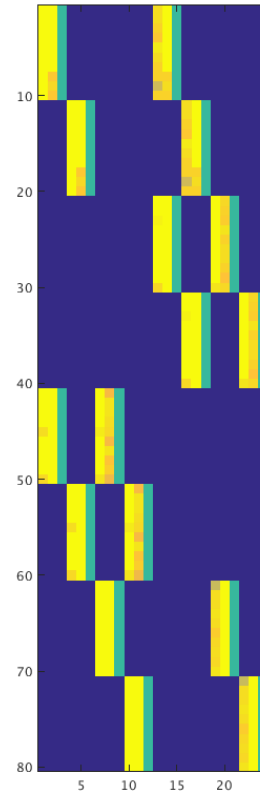
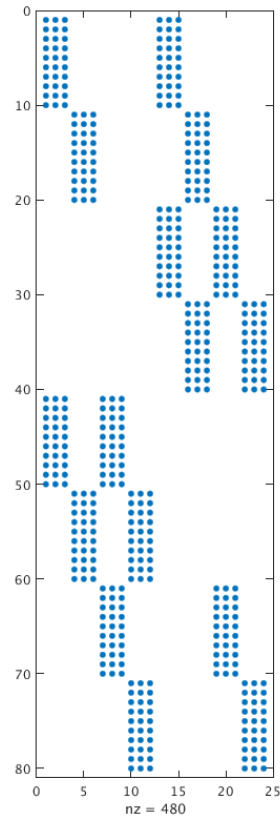
when we expand it to 3 tiles
(with overlaps 1-2, 1-3, but not 2-3)

$$\begin{bmatrix} {}^{1,2}P & - {}^{1,2}Q & 0 \\ {}^{1,3}P & 0 & - {}^{1,3}Q \end{bmatrix} \times \begin{bmatrix} T_1 \\ T_2 \\ T_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

and to 4 tiles
(with overlaps 1-2, 1-3, 2-4, 3-4, but not 1-4, 2-3)

$$\begin{bmatrix} {}^{1,2}P & - {}^{1,2}Q & 0 & 0 \\ {}^{1,3}P & 0 & - {}^{1,3}Q & 0 \\ 0 & {}^{2,4}P & 0 & - {}^{2,4}Q \\ 0 & 0 & {}^{3,4}P & - {}^{3,4}Q \end{bmatrix} \times \begin{bmatrix} T_1 \\ T_2 \\ T_3 \\ T_4 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

Let's cut the point-matches per tile pair down to 10, so it is easier to see this same structure in the example. Below is structure plot and value-coded, to see the repeating structure.



based on the looks of this, the row ordering is a little different than the example above.

It's fine, because we're just re-ordering zeros on the right-hand side.
Instead, we are seeing:

$$\begin{bmatrix} {}^{1,3}P & 0 & -{}^{1,3}Q & 0 \\ 0 & 0 & {}^{3,4}P & -{}^{3,4}Q \\ {}^{1,2}P & -{}^{1,2}Q & 0 & 0 \\ 0 & {}^{2,4}P & 0 & -{}^{2,4}Q \end{bmatrix} \times \begin{bmatrix} T_1 \\ T_2 \\ T_3 \\ T_4 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

(7)

The transforms we are solving for:

$$\begin{bmatrix} T_1 \\ T_2 \\ T_3 \\ T_4 \end{bmatrix}$$

represent 24 unknowns.

The matrix has 80 rows.

The system is overdetermined.

We want a solution for x which minimizes the norm of the residuals:

$$S(x) = ||b - Ax||^2 \quad (8)$$

There are a few forms of this (see wikipedia: normal equations), but, I think this one makes it clear.

A is $m \times n$

$$r_i = b_i - \sum_{j=1}^n A_{ij}x_j \quad (9)$$

$$S = \sum_{i=1}^m r_i^2 \quad (10)$$

minimize with respect to x to find the best fit:

$$\frac{\partial S}{\partial x_j} = 2 \sum_i^m r_i \frac{\partial r_i}{\partial x_j} \quad (11)$$

$$\frac{\partial r_i}{x_j} = -A_{ij} \quad (12)$$

$$\frac{\partial S}{\partial x_j} = 2 \sum_i^m \left(b_i - \sum_{k=1}^n A_{ik} x_k \right) (-A_{ij}) = 0 \quad (13)$$

rearranging:

$$\sum_{i=1}^m \sum_{k=1}^n A_{ij} A_{ik} x_k = \sum_{i=1}^m A_{ij} b_i \quad (14)$$

which is:

$$A^T A x = A^T b \quad (15)$$

$A^T A$ is $n \times n$, so, now we have n equations and n unknowns. The solver includes a way to weight points differently. In this example, this means including an 80×80 diagonal matrix, W :

$$A^T W A x = A^T W b \quad (16)$$

For equal weighting, $W = I$.

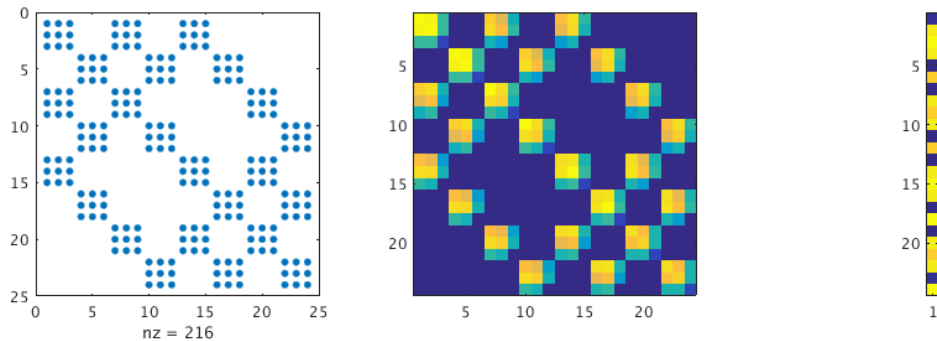
The solver also includes regularization:

$$A^T W A x + \lambda = A^T W b + \lambda d \quad (17)$$

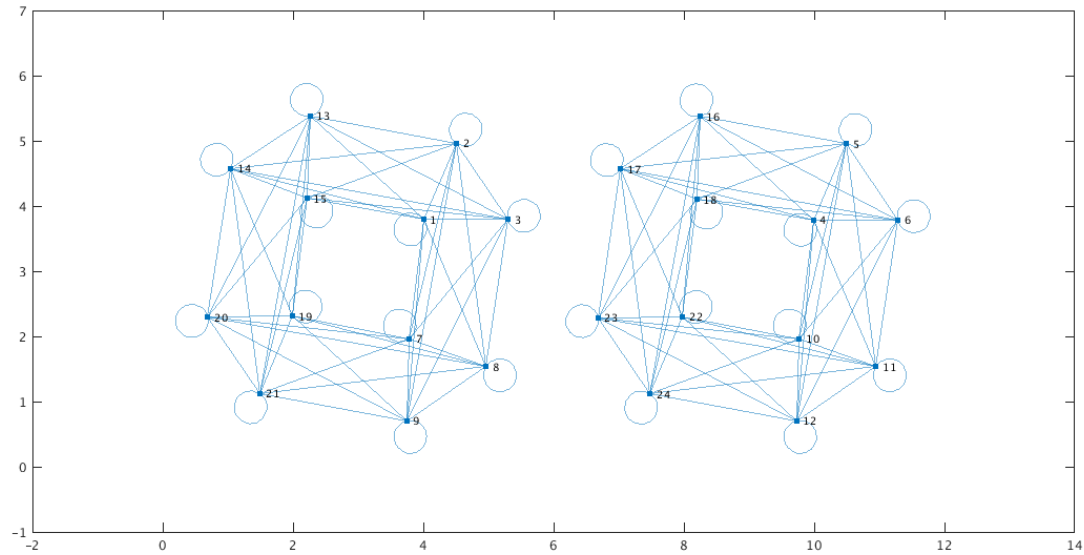
or

$$K x = L_m \quad (18)$$

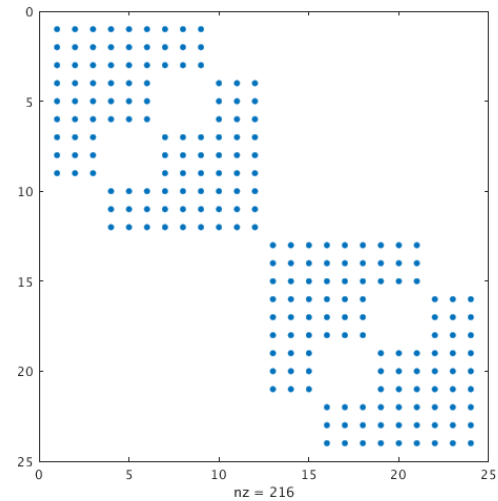
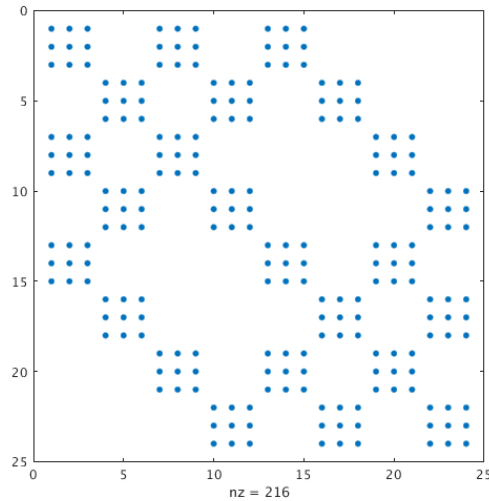
where K is square, and L_m has non-zero entries:



Looking at the graph of K, it shows two disconnected graphs.

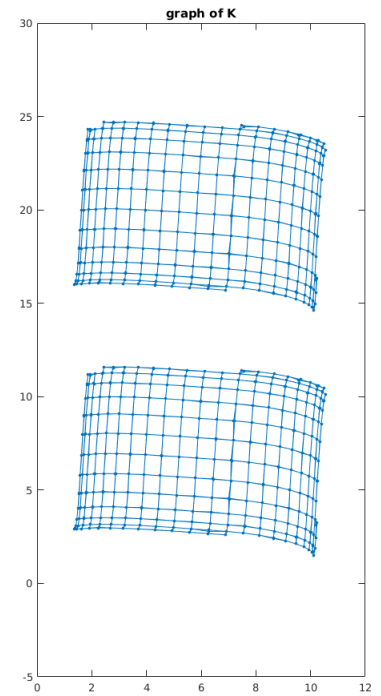
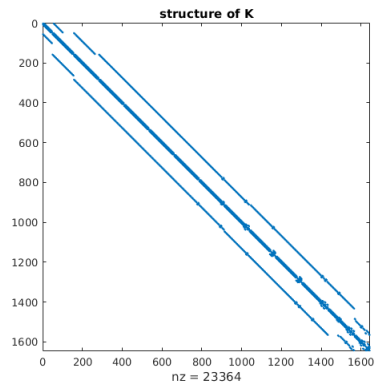
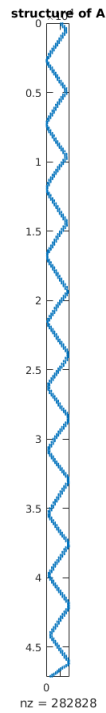


We can use this to permute the rows and columns, to get:

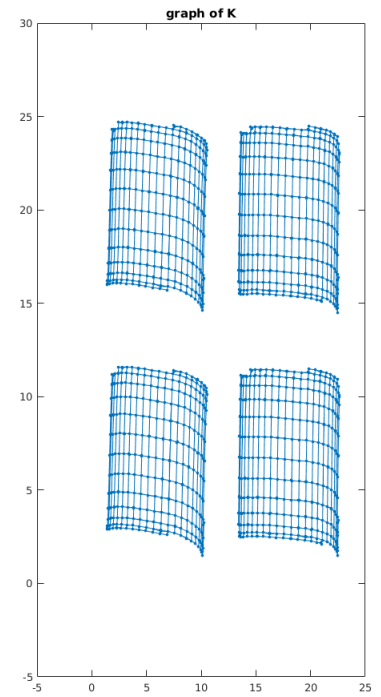
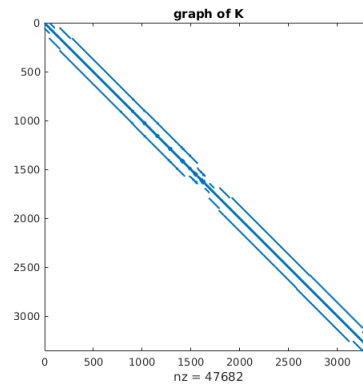
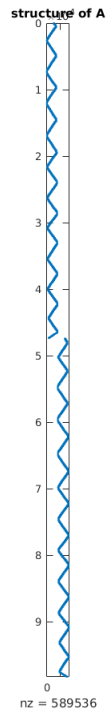


This permuted structure plot implies an obvious way to partition the matrix for distributed solving.

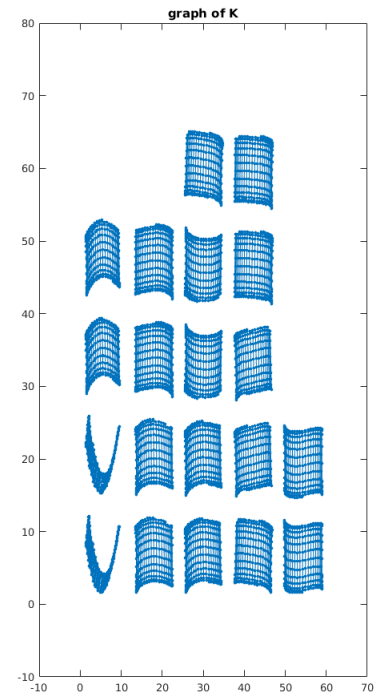
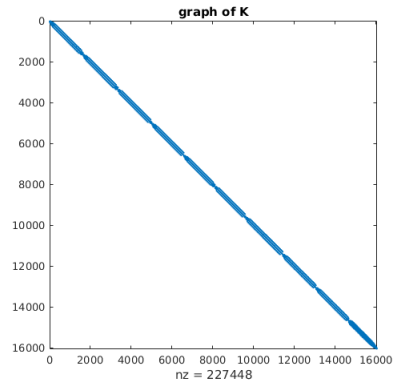
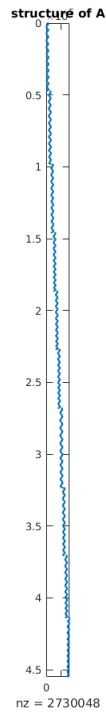
Let's go bigger. A whole section (282 tiles):



now, let's add a second section:



now, 10 sections:



22 sections from EM_Phase1_Fine is the most I can run through right now (unknown code crash):

