

Assignment 1: CS 763

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Problem 1:

The Observed values of average MSE for
q1 is 0.3344
q2 is 0.1795
q3 is 0.6752

which are considerably small compared to the actual value of the x and y coordinates (between 100 and 1000)

The visualization of the points and images can be found in input folder and can be visualized using the **visualize_image.m** script which calls `impixel` info.

Note the images were taken from SL3 where we found the walls with perpendicular surfaces stuck with checkerboard pattern

The reason for normalizing is not clear from the above example as the coordinates are well behaved. If i run it without normalizing and also commenting line 88 then also we get almost the same MSE.

It is a good Idea to Normalize because Then we work with the **same order of magnitude** of the points. this not only helps in **controlling precision errors** but also keeps the numbers small enough to be well **in range** for our computation.

So normalization is essential not only for numerical stability, but also for more accurate estimation in presence of noise and faster solution (in case of iterative solver).

The precise answer as we got from

<https://dsp.stackexchange.com/questions/10887/why-normalize-the-data-set-before-applying-direct-linear-transform?fbclid=IwAR0gwxXtF9ire5mFa13hjV1Cnn7wPyD3-FFtTSMZWdc3uT6pGFOAvpSSH5s>

is that We decrease the condition number of the Matrix M which is the 12*12 matrix we use in the computation according to the code.

In more Detail:-

The source and target coordinate data are usually noisy. Without normalization, the data from source would can have two orders of magnitude larger variance than from target (or vice versa).

The homography estimation usually finds parameters in a least-squares sense - hence the best statistical estimate is found only if variances of the parameters are the same (or known beforehand, but it is more practical just to normalize the input).

Direct solvers do not like poorly scaled problems because numerical instabilities appear (e.g. dividing very large number by a very small number easily leads to numerical overflow).

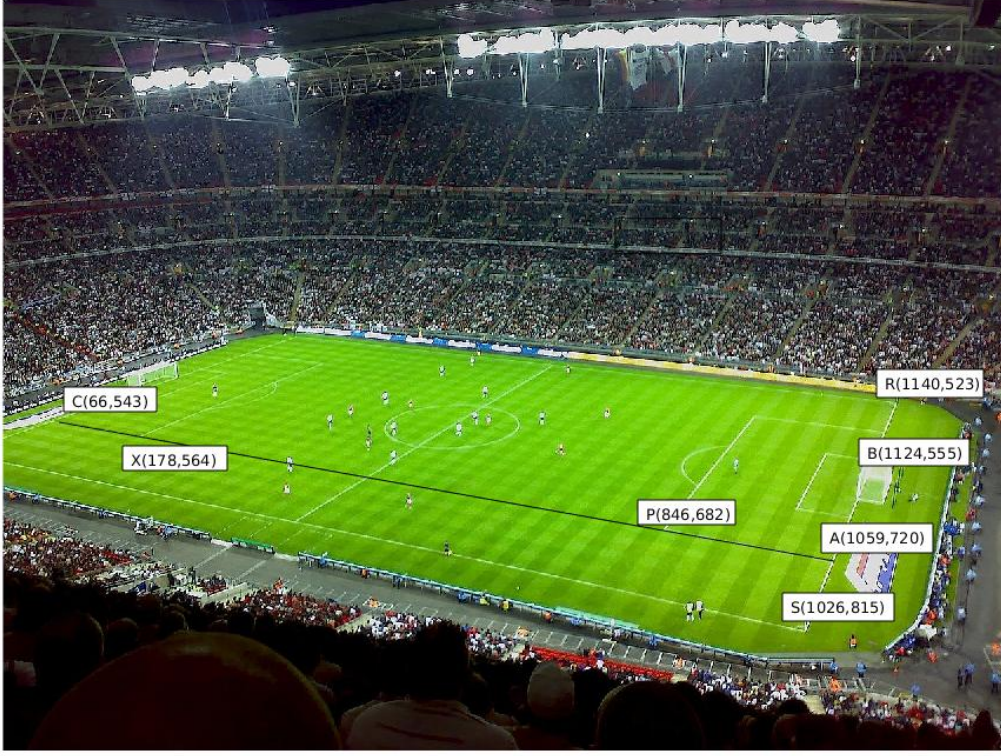
Iterative solvers struggle with badly conditioned matrices by needing more iterations.

Problem 3:

Let G' denote point in the real world corresponding to G in the image, Using cross ratios we have

$$\begin{aligned}\frac{AX * PC}{AC * PX} &= \frac{A'X' * P'C'}{A'C' * P'X'} \\ &= \frac{(A'P' + P'X') * (P'X' + X'C')}{(A'P' + P'X' + X'C') * (P'X')}\end{aligned}$$

Figure 1: Picture with pixel values



Pixel info: (X, Y) [R G B]

Using pixel values from image and using symmetry of football field (ie $A'P' = X'C' = 18$ yards) and replacing $P'C'$ by x we have

$$\begin{aligned} \Rightarrow 1.036 &= \frac{(18 + x) * (x + 18)}{(18 + x + 18) * x} \\ \Rightarrow x &= 78.5 \\ \therefore A'X' &= 18 + x + 18 \\ \Rightarrow A'X' &= 114.5 \quad (\text{length of field}) \end{aligned}$$

Similarly

$$\begin{aligned} \frac{SB * AR}{SR * AB} &= \frac{S'B' * A'R'}{S'R' * A'B'} \\ &= \frac{(S'A' + A'B') * (A'B' + B'R')}{(S'A' + A'B' + B'R') * (A'B')} \end{aligned}$$

Using pixel values from image and using symmetry of football field (ie $S'A' = B'R'$) and replacing $S'A', B'R'$ by x we have

$$\begin{aligned} \Rightarrow 1.07 &= \frac{(x + 44) * (44 + x)}{(44 + 2x) * 44} \\ \Rightarrow x &= 14.7 \\ \therefore S'R' &= x + 44 + x \\ \Rightarrow S'R' &= 73.4 \quad (\text{length of field}) \end{aligned}$$

So, the dimensions of the pitch are 114.5×73.4

The above method uses symmetry of football field (i.e both Dees have same dimensions). The dimensions of the field can also be calculated by calculating the vanishing points along AP and AB (using point of

intersection of parallel lines) and using cross ratios. But the lines in the image are almost parallel so there would be a lot of error in calculating points of intersection. Hence we preferred using symmetry in the field.

Problem 5:

Although there was a high amount of noise in the second image the joint entropy turned out to be minimum when the images were almost aligned.

The joint entropy of an (image, constant image) is same as joint entropy of (image, image). As joint histograms take a straight line shape in both cases.

