

Computer vision (CSC I6716)

Assignment 3

Prof. Zhigang Zhu

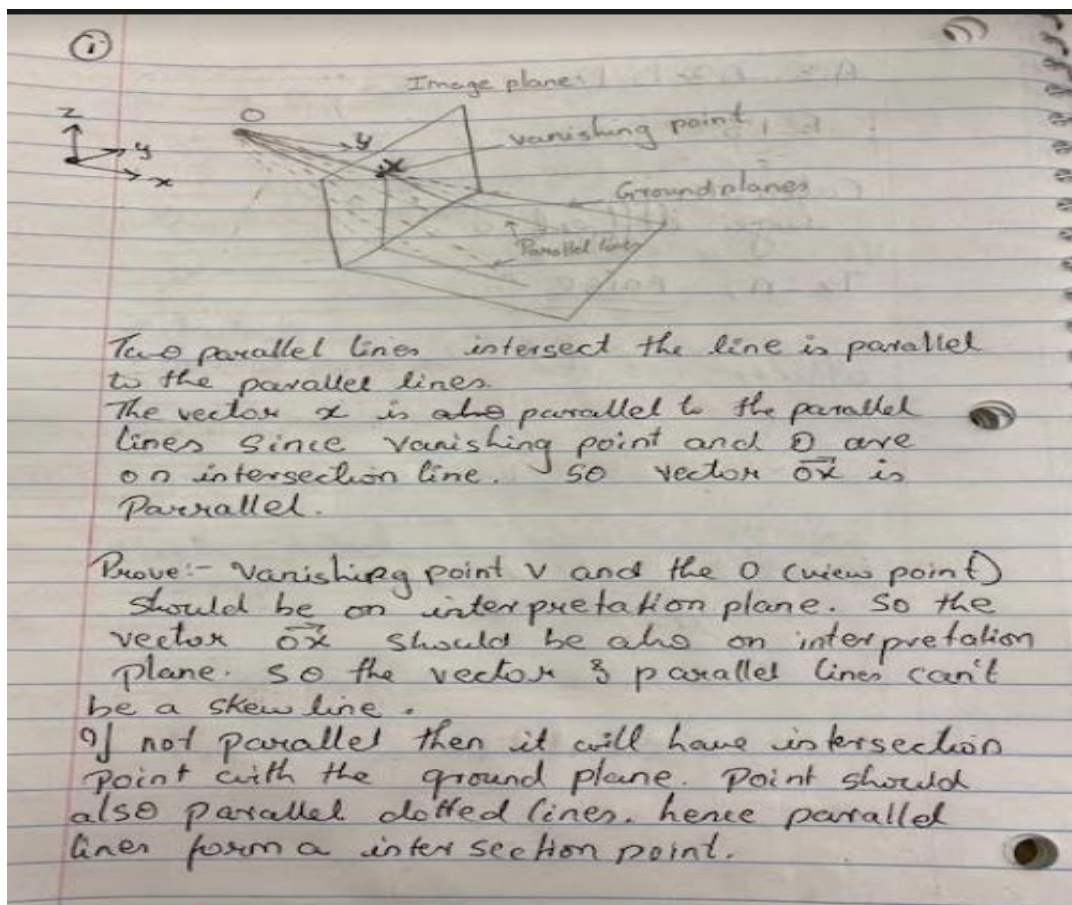
Sonali Shintre

ID: 7602

11/10/19

1. (Camera Models- 30 points) Prove that the vector from the viewpoint of a pinhole camera to the vanishing point (in the image plane) of a set of 3D parallel lines is parallel to the direction of the parallel lines. Please show steps of your proof.

In pinhole camera the focal length varies by given ratio of the image dimension to the distance of the pinhole from the image. Due to perspective, all parallel lines in 3D space appear to meet in a point on the image - the vanishing point, which is the common intersection of all the image lines



If you select to use algebraic calculation, you may use the parametric representation of a 3D line: $P = P_0 + tV$, where $P = (X, Y, Z)^T$ is any point on the line (here T denote for transpose), $P_0 = (X_0, Y_0, Z_0)^T$ is a given fixed point on the line, vector $V = (a, b, c)^T$ represents the direction of the line, and t is the scalar parameter that controls the distance (with sign) between P and P_0 .

Algebraic calculation:-
 3D line : $P = P_0 + tV$
 where $P = (x \ y \ z)^T$
 $P_0 = (x_0 \ y_0 \ z_0)^T$
 vector $V = (a, b, c)^T$
 $t = \text{scalar parameter.}$

Using $x = f(x/z)$ the vanishing point of its image.
 so here $x = P$.

$$\lim_{t \rightarrow \pm\infty} P = f \frac{P_0 + tV}{P_z + tV_z}$$

$$= f \frac{P}{P_z} = f \begin{pmatrix} P_x/P_z \\ P_y/P_z \\ 1 \end{pmatrix}$$

So all the vanishing point meet at the parallel lines.

2. (Camera Models- 20 points) Show that relation between any image point $(x_{im}, y_{im})^T$ of a plane (in the form of $(x_1, x_2, x_3)^T$ in projective space) and its corresponding point $(X_w, Y_w, Z_w)^T$ on the plane in 3D space can be represented by a 3×3 matrix. You should start from the general form of the camera model $(x_1, x_2, x_3)^T = M_{int} M_{ext} (X_w, Y_w, Z_w, 1)^T$, where the image center (o_x, o_y) , the focal length f , the scaling factors $(s_x \text{ and } s_y)$, the rotation matrix R and the translation vector T are all unknown. Note that in the course slides and the lecture notes, I used a simplified model of the perspective project by assuming o_x and o_y are known and $s_x = s_y = 1$, and only discussed the special cases of planes. So you cannot directly copy those equations I used. Instead you should use the general form of the projective matrix, and the general form of a plane $n_x X_w + n_y Y_w + n_z Z_w = d$.

(2)

Plane eqn

$$n_x x_w + n_y y_w + n_z z_w = d$$

$$= n^T P_w = d$$

P_w is a 3D point on the plane $= (x_w, y_w, z_w)^T$

z_w can be rewritten as a function of x_w & y_w

$$\therefore z_w = 0$$

$$\therefore P_w = (x_w, y_w, 0)^T$$

3D point (x_w, y_w, z_w) becomes a 2D point (x_w, y_w)

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{bmatrix} -f_{y11} & -f_{y12} & -f_{Tx} \\ -f_{y21} & -f_{y22} & -f_{Ty} \\ r_{31} & r_{32} & T_z \end{bmatrix} \begin{pmatrix} x_w \\ y_w \\ 1 \end{pmatrix}$$

$n_z = 1$ so plane eqn will be

$$z_w = d - n_x x_w - n_y y_w$$

$$H = \begin{bmatrix} -f_{y11} & -f_{y12} & -f_{y13} & -f_{Tx} \\ -f_{y21} & -f_{y22} & -f_{y23} & -f_{Ty} \\ r_{31} & r_{32} & r_{33} & T_z \end{bmatrix}$$

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{bmatrix} -f(y_{11} - n_x y_{13}) - f(y_{12} - n_y y_{13}) - f(dy_{13} + T_x) \\ -f(y_{21} - n_x y_{23}) - f(y_{22} - n_y y_{23}) - f(dy_{23} + T_y) \\ (r_{31} - n_x r_{33}) & (r_{32} - n_y r_{33}) & (dr_{33} + T_z) \end{bmatrix} \begin{pmatrix} x_w \\ y_w \\ 1 \end{pmatrix}$$

$$M_{ext} = \begin{bmatrix} \gamma_{11} & \gamma_{12} & \gamma_{13} & T_x \\ \gamma_{21} & \gamma_{22} & \gamma_{23} & T_y \\ \gamma_{31} & \gamma_{32} & \gamma_{33} & T_z \end{bmatrix} = \begin{bmatrix} R_1^T & T_x \\ R_2^T & T_y \\ R_3^T & T_z \end{bmatrix}$$

M_{ext} = includes all extrinsic parameters
 M_{int} = all intrinsic parameters

$$M_{int} = \begin{bmatrix} -f_x & 0 & 0_x \\ 0 & -f_y & 0_y \\ 0 & 0 & 1 \end{bmatrix}$$

Combine M_{ext} and M_{int} in eq 1

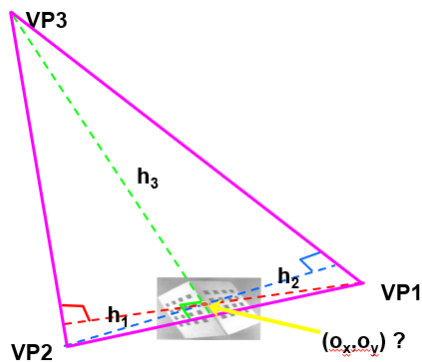
$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = M_{int} M_{ext} \begin{pmatrix} x_w \\ y_w \\ z_w \\ 1 \end{pmatrix}$$

Each side divide by x_3

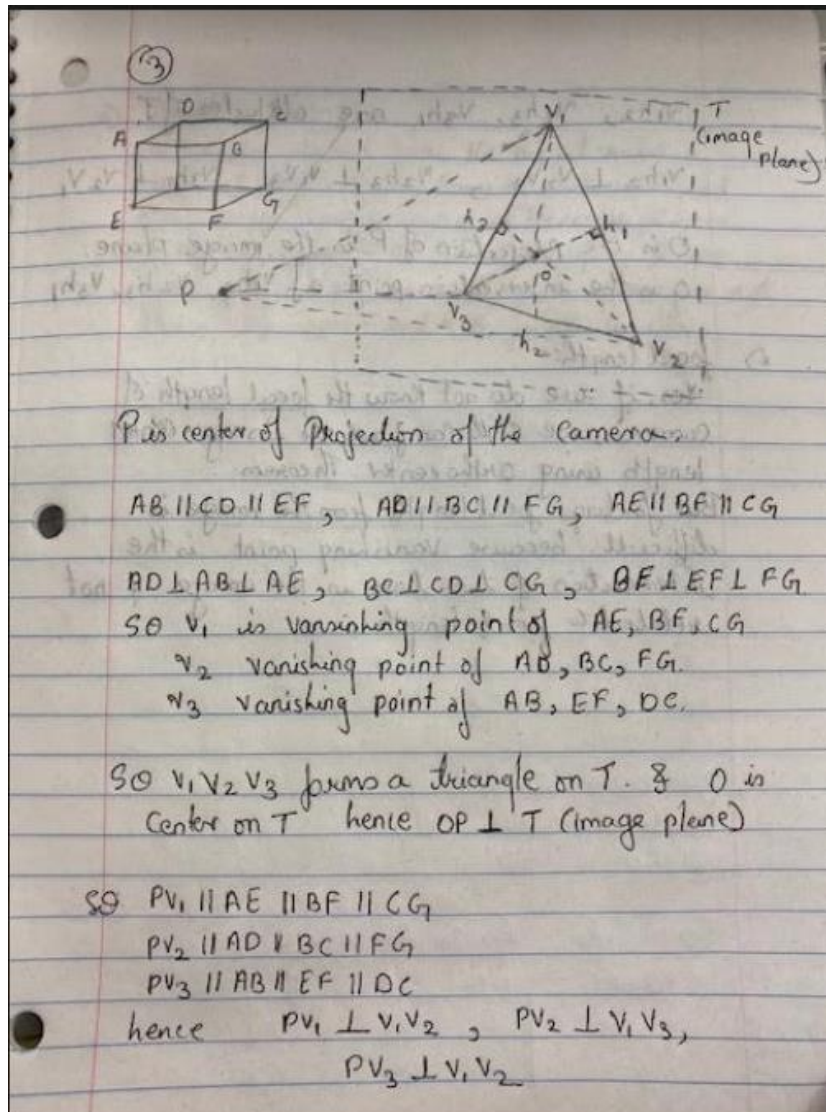
$$\begin{pmatrix} x_{im} \\ y_{im} \end{pmatrix} = \begin{pmatrix} x_1/x_3 \\ x_2/x_3 \end{pmatrix}$$

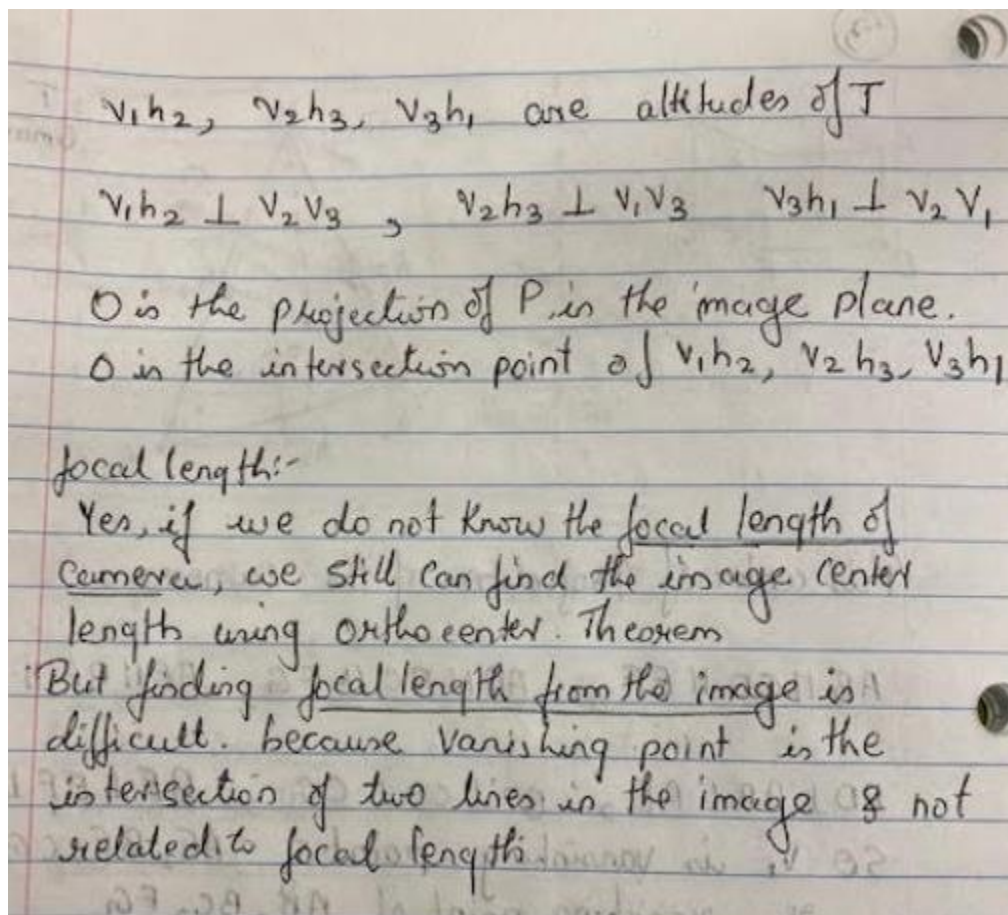
3. (Calibration- 20 points) Prove the Orthocenter Theorem by geometric arguments: Let T be the triangle on the image plane defined by the three vanishing points of three mutually orthogonal sets of parallel lines in space. Then the image center is the orthocenter of the triangle T (i.e., the common intersection of the three altitudes).
- (1) Basic proof: use the result of Question 1, assuming the aspect ratio of the camera is 1. (10 points)
 - (2) If you do not know the focal length of the camera, can you still find the image center (together with the focal length) using the Orthocenter Theorem? Show why or why not. (5 points)
 - (3) If you do not know the aspect ratio and the focal length of the camera, can you still find the image center using the Orthocenter Theorem? Show why or why not. (5 points)

The center of projection of the camera in 3D space is $O(x, y)$. Three mutually orthogonal sets of parallel lines: L_1, L_2 , and L_3 . Assume there is a triangle formed by lines $V_1V_2V_3$, and then they are also V_1, V_2 , and V_3 are the three vanishing points.



i. proof:





ii. Focal Length:

Yes, if we do not know the focal length of the camera we can still find the image center length using orthocenter theorem, But finding focal length from the image is difficult, since vanishing point is intersection of the two lines in the image and not related to focal length

iii. Aspect Ratio:

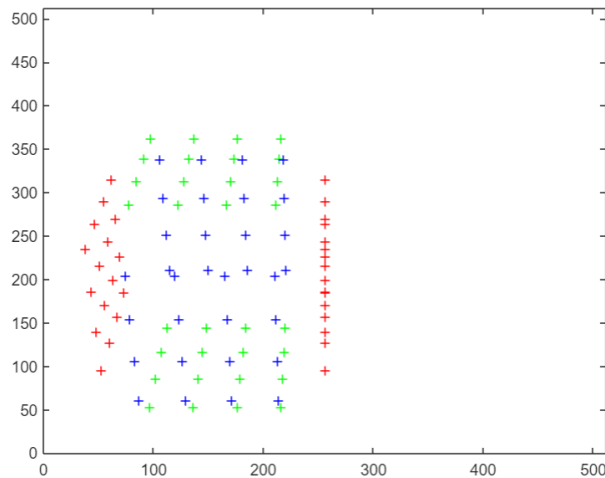
No if we do not know the aspect ratio and the focal length of the camera, we cannot find the image center using orthocenter theorem, since vanishing point is not unique and we can't get the unique orthocenter if we do not know the aspect ratio and focal length

4. **Calibration Programming Exercises (40 points):** Implement the direct parameter calibration method in order to (1) learn how to use SVD to solve systems of linear equations; (2) understand the physical constraints of the camera parameters; and (3) understand important issues related to calibration, such as calibration pattern design, point localization accuracy and robustness of the algorithms. Since calibrating a real camera involves lots of work in calibration pattern design, image processing and error controls as well as solving the equations, we will mainly use simulated data to understand the algorithms. As a by-product we will also learn how to generate 2D images from 3D models using a "virtual" pinhole camera.

1. Generated a virtual 3D cube. as a result, a graph is generated and, on the graph, at each surface there are 4x4 points. I need to use those points as variables for an equation.

$X_w=0, Y_w=0, X_w=1, Y_w=1, Z_w=1, Z_w=0$. so here P_w is a $1 \times 1 \times 1$ m3 cube, 96 points are generated and each surface have 16 points.

Total 96 points for calibration



2. create virtual camera with known intrinsic parameters given focal length $f=16\text{mm}$, image center $(o_x, o_y) = (256, 256)$, and pixel size (s_x, s_y) . Using three rotation angles α, β, γ to generate the rotation matrix R , I will tilt it to 30 degree to have better projection.

4) So Total 96 points used for Test

- 1) calibration
- 2) Virtual camera :- intrinsic parameter

$f = 0.016$
 $o_x = 256$
 $o_y = 256$
 $s_x = 0.0088/5120$
 $s_y = 0.0066/512$

$F_x = f/s_x = 930.90$
 $F_y = 1241.2$

$\arg = \frac{F_x}{F_y} \approx 0.75$

$T =$	0	0	1
	T_x	T_y	T_z

$\gamma = 0^\circ$
 $\beta = 0^\circ$
 $\alpha = 30^\circ$

So :- $R = R_\alpha \times R_\beta \times R_\gamma$

Matrix =

1.0	0	0
0	0.86	-0.5
0	0.5	0.8

3. Estimating the intrinsic (f_x, f_y , aspect ratio a , image center (o_x, o_y)) and extrinsic (R, T and further alpha, beta, gamma)

- I. first, we know the image center $O_x = 256; O_y = 256$, every point in 3D have corresponding point image, first we find the matrix A where $A = UDV^T$. If the matrix A is a real matrix, then U and V are also real. Calculate T_z, F_x, F_y , forming 8 cols and 96 rows, which 8 unknowns $v = (v_1, \dots, v_8)^T$. 7 of them are independent parameters and 8th row is zero.

$$x_i X_i v_1 + x_i Y_i v_2 + x_i Z_i v_3 + x_i v_4 - y_i X_i v_5 - y_i Y_i v_6 - y_i Z_i v_7 - y_i v_8 = 0$$

$$(v_1, v_2, v_3, v_4, v_5, v_6, v_7, v_8) = (r_{21}, r_{22}, r_{23}, T_y, cr_{11}, cr_{12}, cr_{13}, \alpha T_x)$$

$$Av = 0$$

3) $O_x = 256$ $O_y = 256$

i) Eq: $x_i X_i v_1 + x_i Y_i v_2 + \dots$
 with 96 data points we have 8 unknowns
 8 cols and 96 rows.

eq:-
 $x_i X_i v_1 + x_i Y_i v_2 + x_i Z_i v_3 + x_i v_4 - y_i X_i v_5 - y_i Y_i v_6 - y_i Z_i v_7 - y_i v_8 = 0$

$Av = 0$

A matrix $[96 \times 8]$

find $V = [u, s, v] = \text{svd}(A)$;
 (8×8)

$v = [1 \times 8]$

$\alpha = \sqrt{v_5^2 + v_6^2 + v_7^2}$

$\alpha = \sqrt{v_5^2 + v_6^2 + v_7^2} / \sqrt{v_1^2 + v_2^2 + v_3^2}$

$\alpha = 4.62 / 2.88 = 1.6$

A matrix =

-217.8480	-43.5696	-43.5696	-217.8480	21.2634	4.2527	4.2527	21.2634
-209.3619	-41.8724	-83.7448	-209.3619	-7.4798	-1.4960	-2.9919	-7.4798
-201.5123	-40.3025	-120.9074	-201.5123	-34.0676	-6.8135	-20.4406	-34.0676
-194.2299	-38.8460	-155.3839	-194.2299	-58.7338	-11.7468	-46.9870	-58.7338
-212.8666	-85.1466	-42.5733	-212.8666	69.9366	27.9747	13.9873	69.9366
-204.7570	-81.9028	-81.9028	-204.7570	39.9713	15.9885	15.9885	39.9713
-197.2426	-78.8970	-118.3455	-197.2426	12.2054	4.8822	7.3232	12.2054
-190.2602	-76.1041	-152.2082	-190.2602	-13.5947	-5.4379	-10.8757	-13.5947
-208.1078	-124.8647	-41.6216	-208.1078	116.4336	69.8602	23.2867	116.4336

.....

SVD: [U,V,S]=svd(A)

V:

0.4618	-0.3469	0.3377	-0.0749	0.3104	-0.0129	0.6710	-0.0000
0.3374	0.0467	-0.0292	-0.4083	-0.4777	0.0892	-0.0163	0.6928
0.2443	-0.3095	-0.6967	-0.2771	-0.0977	0.3332	0.0432	-0.4000
0.5662	-0.3648	0.1026	0.3206	0.0738	-0.1695	-0.6316	0.0000
-0.2268	-0.2603	-0.4307	0.2868	0.4865	0.1192	0.0476	0.6000
-0.3162	-0.3905	0.4293	-0.4108	0.1704	0.5159	-0.3152	0.0000
-0.0990	-0.2486	0.1312	0.6121	-0.5509	0.4369	0.2052	0.0000
-0.3626	-0.6055	-0.0470	-0.1497	-0.3021	-0.6172	0.0714	-0.0000

v= [-0.0000 0.6928 -0.4000 0.0000 0.6000 0.0000 0.0000 -0.0000]

$$|\gamma| = \sqrt{\bar{v}_1^2 + \bar{v}_2^2 + \bar{v}_3^2}$$

$$\alpha = \sqrt{\bar{v}_5^2 + \bar{v}_6^2 + \bar{v}_7^2} / \sqrt{\bar{v}_1^2 + \bar{v}_2^2 + \bar{v}_3^2}$$

Calculating Fx, Fy

Calculating aspect

aspect ratio	ground true	Estimated	diff
act	0.75	0.7544	0.0044

T_z, F_x, F_y

F_x, F_y

	Ground True	Estimated	diff
F_x	930.9	930.9	0
F_y	1241.2	1241.2	0

Calculate $R \& T$

$$R_3^T = R_1^T \times R_2^T = SR_1^T \times SR_2^T$$

Comparing ground True & Estimated R is identical to the virtual camera.

Estimated R_c

$$\begin{matrix} -1.0 & 0.0 & 0.0 \\ 0.0 & -0.8660 & 0.5000 \\ -0.0 & 0.50 & 0.8660 \end{matrix}$$

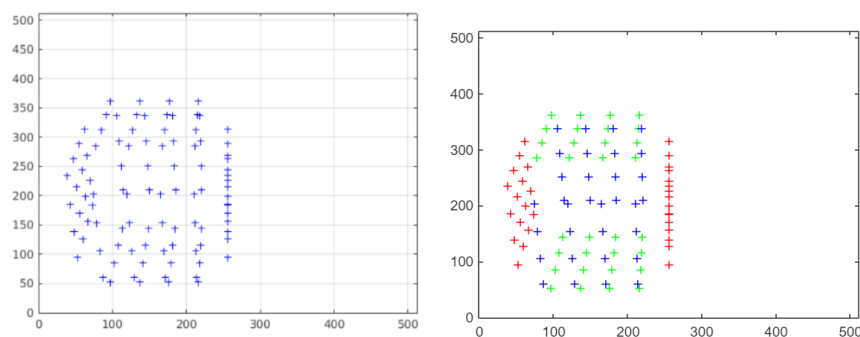
Calculate $T = [0, 0, 4]$.

	Ground True	Estimated values	diff
T_x	0	$6.7009e-04$	$6.7009e-04$
T_y	0	$1.8468e-16$	$1.8468e-16$
T_z	4	4.	4.

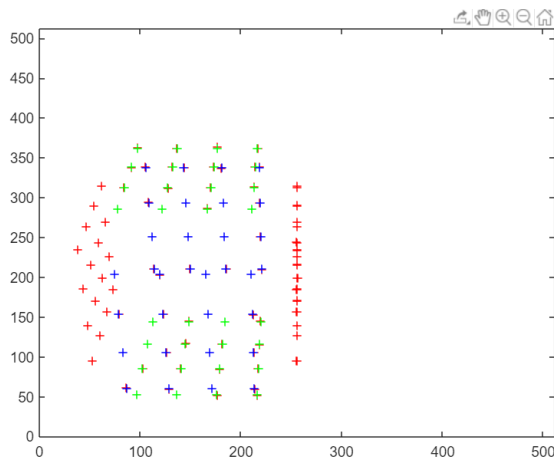
ii) After calculating T_x, T_y, T_z , aspect ratio, f_x, f_y from the ground True & Estimated. We see that only T_x, T_y influence the image center. So we should have fixed image center when using direct parameter calibration.

- ii. If the ratio of an image is different than the ratio of your screen, we may not see the whole image. Images won't fit properly if the screen is narrower than the image. We know the first 2 rows of R and first two components from the T. By this 3rd row will calculate from the R1 and R2. Next is to Determine the sign s positive or negative, assuming s as positive, test sign of s use the image point. We got $X = -1$ and $x = 1$ so s is negative in this case.

if we do not know the aspect ratio of the camera, we can't find the image center using Orthocenter Theorem. Because the vanishing point is not unique, and we can't get the unique orthocenter if we do not know the aspect ratio.



- iii. Adding some noise, 0.1 mm random error to 3D points and 0.5 pixel random error to 2D points. Once noise is added we can Then we can constraints by using SVD.



iii) Random noise:-

After adding 0.1 mm random error to 3D points
 & 0.5 pixels random error to 2D points.

	Ground True	noise add	diff
aspect ratio	0.75	0.7491	0.0009
Fx	980.9	986.9095	6.9905
Fy	1241.2	1242.7	1.5
Tx	$6.7009E-0.4$	$-2.6265E-0.4$	0.00093
Ty	$1.8468E-16$	$-6.5654E-0.4$	0.00055
Tz	4.	3.9453	0.05

As the result
 aspect ratio, Fx, Fy, Ty, Tz, Tx are all
 changed by adding noise.

We can conclude that orthocenter method
 is very sensitive to the extrinsic
 parameter.

WE can conclude that orthocenter method is sensitive to the extrinsic parameter.

