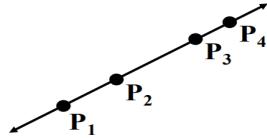


Question 1

Cross ratio

(a) Prove that the cross-ratio is invariant under the projective transformation (use homogeneous coordinates):
cross-ratio means that a straight line under a projective transformation remains straight.

And it's supposed to be invariant accross projective transformations:



$$\frac{\|\mathbf{P}_3 - \mathbf{P}_1\| \|\mathbf{P}_4 - \mathbf{P}_2\|}{\|\mathbf{P}_3 - \mathbf{P}_2\| \|\mathbf{P}_4 - \mathbf{P}_1\|} \quad \mathbf{P}_i = \begin{bmatrix} X_i \\ Y_i \\ Z_i \\ 1 \end{bmatrix}$$

Let A, B, C, D be 4 collinear points (on the same line) and let H be a projective transformation.

We know from the property of projective transformations that they preserve straightness and so $HA, HB, HC, HD = A', B', C', D'$

are on the same line.

Lets denote another point $O = \begin{pmatrix} o_1 & o_2 & 1 \end{pmatrix}$ and $HO = O'$ a point that is not collinear with A', B', C', D'

Then we get that $(A', B'; C', D') = \frac{[O', A', C'] \cdot [O', B', D']}{[O', B', C'] \cdot [O', A', D']}$ from Menelaus' theorem

It also holds that

$$[O', A', C'] = \begin{vmatrix} O'_1 & A'_1 & C'_1 \\ O'_2 & A'_2 & C'_2 \\ 1 & 1 & 1 \end{vmatrix} = \begin{vmatrix} O'_1 & A'_1 - O'_1 & C'_1 - O'_1 \\ O'_2 & A'_2 - O'_2 & C'_2 - O'_2 \\ 1 & 0 & 0 \end{vmatrix} = 1 \cdot \begin{vmatrix} A'_1 - O'_1 & C'_1 - O'_1 \\ A'_2 - O'_2 & C'_2 - O'_2 \end{vmatrix} = [HOHA, HOHC] = \det(HOH) \cdot [A, C]$$

It can be done to $[O', B', D']$, $[O', B', C']$, $[O', A', D']$ as well:

$$\text{So } (A', B'; C', D') = \frac{[O', A', C'] \cdot [O', B', D']}{[O', B', C'] \cdot [O', A', D']} = \frac{\det(HOH)^2 [A, C] \cdot [B, D]}{\det(HOH)^2 [B, C] \cdot [A, D]} = \frac{[A, C] \cdot [B, D]}{[B, C] \cdot [A, D]} = (A, B; C, D)$$

(b) Measure size of table

$d = ab \times a'b'$ (a point in 'infinity')



$T = \text{table size}$

$S = \text{size of the book}$

I used this code for calculations:

$$\frac{\|c-a\|\|d-b\|}{\|b-a\|\|d-c\|} = \frac{T}{S}$$
$$\Rightarrow S \cdot \frac{\|c-a\|\|d-b\|}{\|b-a\|\|d-c\|} = T$$

```
5     REAL_DIST_AB = 25.5
6     def run_compute_table_size():
7         a = np.array([17724, 2863, 1])
8         b = np.array([1310, 1743, 1])
9         a_dot_b = np.array([3964, 1792, 1])
10        b_dot_a = np.array([2363, 1299, 1])
11        c = np.array([166, 857, 1])
12        line_ab = do_cross(a, b)
13        line_atb = do_cross(a_dot_b, b_dot_a)
14        d = do_cross(line_atb, line_ab)
15        T = dist(c, a) * dist(d, b)/(dist(b, a) * dist(d, c)) * REAL_DIST_AB
16        print("size of table is [np.round(T, 1)] cm, vs real size of 110 cm")
17    def dist(x, y):
18        return np.linalg.norm(x[-1] - y[-1])
19    def do_cross(a, b):
20        line_ab = np.cross(a, b).astype(float)
21        line_ab /= line_ab[-1]
22        return line_ab
23
24
def()
```

```
main()
"C:\Users\anita\OneDrive\Desktop\master\vision 3D\ex2\venv\Scripts\python.exe" "C:\
```

```
size of table is 114.2 cm, vs real size of 110 cm
```

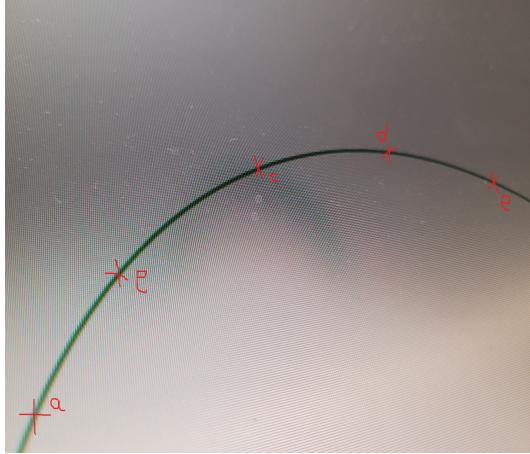
The result I got was 3.5 percent of the real distance.

Question 2

I found 6 points on a circle that was picured from the middle.

fit the points to a conic and found the conic parameters using SVD and the following code:

The discriminant came extremely small and this is why I can say that it was indeed projected to the image as a parabula.



I used this code: created the conic points matrix A, and then found the nullspace of this matrix the represented the conic variables.

because the data is noisy there was neber going to be a perfect parabula, and therefore thee discriminant was never going to be exactly 0

and so I wanted to be smaller then a certain ε in order to state that this is indeed a parabula. And iit came out $1.08 \cdot 10^{-14}$ which is very small.

```
def analyze_circle_parabula(points):
    A = np.zeros((len(points), 6))
    for i, (x, y) in enumerate(points):
        A[i] = [x ** 2, x * y, y ** 2, x, y, 1]
    U, S, V = np.linalg.svd(A)
    a, b, c, d, e, f = V[-1]
    print(f"Conic equation: {a}x^2 + {b}xy + {c}y^2 + {d}x + {e}y + {f} = 0")
    discriminant = b ** 2 - 4 * a * c
    print(f"Discriminant is {discriminant}")
    if np.abs(discriminant) < 1e-10:
        print("this conic is a parabula")

if __name__ == "__main__":
    a = [251, 2723]
    b = [875, 1084]
    c = [1935, 1229]
    d = [2923, 1135]
    e = [3717, 1321]
    f = [1582, 1497]
    points = np.array([a, b, c, d, e, f])
    f_name_ = __file__
    f_name_ = f_name_.split('.')[0]
    with open(f_name_ + ".inp", "w") as f:
        f.write(str(points))
    main()
    print(f"Discriminant is {discriminant}")
    print(f"this conic is a parabula")
```

Question 3

Extracting camera location:



I used the following points:

the real world coordinates are:

$$a = (60, 0, 0, 1) \quad b = (0, 0, 0, 1) \quad c = (60, 60, 0, 1)$$

$$d = (0, 60, 0, 1) \quad e = (0, 105, 10, 1) \quad f = (60, 105, 10, 1)$$

And the image coordinates are:

$$A = (1862, 2570, 1) \quad B = (162, 2097, 1) \quad C = (1988, 1511, 1)$$

$$D = (880, 1379, 1) \quad E = (1137, 984, 1) \quad F = (2049, 1022, 1)$$

And use the following points to find the elements of the camera matrix using the following formula:

$$\begin{bmatrix} X_1 & Y_1 & Z_1 & 1 & 0 & 0 & 0 & -u_1X_1 & -u_1Y_1 & -u_1Z_1 & -u_1 \\ 0 & 0 & 0 & 0 & X_1 & Y_1 & Z_1 & 1 & -v_1X_1 & -v_1Y_1 & -v_1Z_1 & -v_1 \\ X_n & Y_n & Z_n & 1 & 0 & 0 & 0 & -u_nX_n & -u_nY_n & -u_nZ_n & -u_n \\ 0 & 0 & 0 & 0 & X_n & Y_n & Z_n & 1 & -v_nX_n & -v_nY_n & -v_nZ_n & -v_n \end{bmatrix} \begin{bmatrix} m_{00} \\ m_{01} \\ m_{02} \\ m_{03} \\ m_{10} \\ m_{11} \\ m_{12} \\ m_{13} \\ m_{20} \\ m_{21} \\ m_{22} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 0 \end{bmatrix}$$

I found the camera matrix using this code:

basically it's building the big matrix from the above formula, and then finds its nullspace using the last column of the SVD decomposition.

```
def find_fundamental_matrix(world_points, image_points):
    n = world_points.shape[0]
    # Construct the matrix A
    A = np.zeros((2 * n, 12))
    for i in range(n):
        x, y, z = world_points[i][-1]
        u, v = image_points[i][-1]
        A[2 * i] = [x, y, z, 1, 0, 0, 0, 0, -u * x, -u * y, -u * z, -u]
        A[2 * i + 1] = [0, 0, 0, 0, x, y, z, 1, -v * x, -v * y, -v * z, -v]
    # Solve for the nullspace of A
    _, _, V = np.linalg.svd(A)
    P = V[-1].reshape(3, 4)
    P /= P[-1, -1]
    return P
```

Then I found the C component:

$$\mathbf{P} = \mathbf{K}\mathbf{R}[\mathbf{I}] - \mathbf{C}$$

↑ ↑ ↑ ↑
3x3 3x3 3x3 3x1
intrinsic 3D rotation identity 3D translation

Using this code:

```
# Compute the camera center
C = -np.linalg.inv(P[:, :3]) @ P[:, 3]
```

The result that I got was:

```
camera location is [ 59.06059393 -36.792746  58.83892228] cm
```

which make sense as $b=(0,0,0)$ and $a=(60, 0, 0)$

So it's in a's x axis, negative y and z around 60 cm.