ECE 417 Midterm 2 2022 practice problem set

Instructor: Vikas Dhiman March 30th, 2021

(1) Student name:

Student email:

About the exam

- 1. There are total 5 problems. You must attempt all 5.
- 2. Maximum marks: 50 (70 with bonus marks).
- 3. Maximum time allotted: 50 min
- 4. Calculators are allowed.
- 5. One US Letter size or A4 size cheat sheet (both-sides) is allowed.

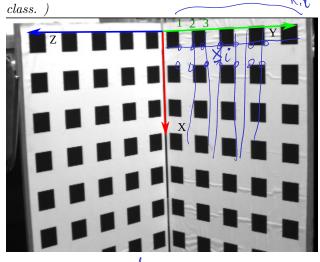
$$x_{76} = \begin{pmatrix} 7_{1} & 3_{-2} & 3_{-2} \\ x_{1} & 2_{1} \end{pmatrix}$$

$$x_{1} = \begin{pmatrix} x_{1} \\ x_{1} \\ x_{1} \end{pmatrix}$$

Problem 1 Given a set of $n \geq 6$ points $\underline{\mathbf{X}}_i \subseteq \mathbb{P}^3$ for all $i \in \{1, \dots, n\}$ in 3D projective space, and a set of corresponding points $\underline{\mathbf{u}}_i \in \mathbb{P}^2$ in an image, find the 3D to 2D projective $P \in \mathbb{R}^{3 \times 4}$ matrix that converts \mathbf{X}_i to $\underline{\mathbf{u}}_i = \lambda_i P \underline{\mathbf{X}}_i$. In other words, convert $\underline{\mathbf{u}}_i \times P \underline{\mathbf{X}}_i = 0$ into a familiar form $A\mathbf{y} = \mathbf{b}$ or $A\mathbf{y} = \mathbf{0}$ so that we can

 $\underline{\mathbf{u}}_i = \lambda_i P \underline{\mathbf{X}}_i$. In other words, convert $\underline{\mathbf{u}}_i \times \mathbf{r} \underline{\mathbf{A}}_i - \mathbf{0}$ and $\mathbf{u}_i = [x_i, y_i, w_i]^{\top}$ and $P = \begin{bmatrix} \mathbf{p}_1^{\top} \\ \mathbf{p}_2^{\top} \\ \mathbf{p}_3^{\top} \end{bmatrix}$ where $\mathbf{p}_1, \mathbf{p}_2, \mathbf{p}_3 \in \mathbb{R}^4$

are the rows of P represented as 4-D column vectors. (Practical motivation: We did camera calibration in lab using a single checker board. It is much easier to compute camera calibration using two mutually perpendicular checker boards so that all points do not lie on a single plane (hence linearly independent). One can make a coordinate system attached to the double checker and compute the 3D coordinates of each corner point in that system. Let $\underline{\mathbf{X}}_i \in \mathbb{P}^3$ be such points in 3D on the checker-board. Let $\underline{\mathbf{u}}_i \in \mathbb{P}^2$ be a point detected in the image so that we have one-to-one correspondence between $\underline{\mathbf{X}}_i$ and $\underline{\mathbf{u}}_i$. Finding the projection matrix $P \in \mathbb{R}^{3 \times 4}$ then reduces to the above problem. We will cover the breakdown of P matrix into P = K[R,t] in



PER³×4

Wi =

known

GixPXi = 0

P1/P2/P3 EIR 4×1 vectors

 $u_{i} = \lambda_{i} K(RX_{i} + t)$ $u_{i} = \lambda_{i} K(RX_{i} + t)$ $u_{i} = \lambda_{i} PX_{i}$ $u_{i} = \lambda_{i} PX_{i}$

 $u_i = \lambda_1 P X_1$, $u_2 = \lambda_2 P X_2$ -- _

unz InPXn

 $\lambda_i \in \mathbb{R}$

Ui sxi vedo

$$\frac{2}{\sqrt{2}} = \frac{1}{\sqrt{2}} \times \frac{1}{\sqrt{2}} = \frac{1$$

 $\frac{\alpha \| b + \alpha \times b = 0}{\alpha \times b \| = \|a\| \|b\| \le m\theta}$

$$u_i \times = \begin{pmatrix} 0 & -\omega_i & y_i \\ \omega_i & 0 & -x_i \\ -y_i & x_i \end{pmatrix}$$

$$PXi = 0$$

$$PXi$$

$$(y_{1})_{x}PX_{i} = O_{3x},$$

$$(y_{1})_{x}PX_{i} = O_{3x},$$

$$(y_{1})_{x}PX_{i}P_{2}$$

$$(y_{2})_{x}Y_{i}P_{2}$$

$$(y_{3})_{x}PX_{i}P_{2}$$

$$(y_{1})_{x}PX_{i}P_{2}$$

$$(y_{2})_{x}PX_{i}P_{2}$$

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$$(y_{3})_{x}PX_{i}PX_{i}P_{2}$$

$$(y_{3})_{x}PX_{i}PX_{$$

$$\begin{pmatrix}
O_{1x1}^{T} & -\omega_{1}X_{1}^{T} & y_{1}X_{1}^{T} \\
\omega_{1}X_{1}^{T} & O_{1x1}^{T} & -\lambda_{1}X_{1}^{T}
\end{pmatrix}
\begin{pmatrix}
b_{1} \\
b_{2} \\
b_{3}
\end{pmatrix}_{12x1}$$

$$\begin{pmatrix}
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O_{1x1}^{T} & -\lambda_{1}X_{1}$$

> yema)

$$A = \bigcup_{2n \times 2n} \bigvee_{12 \times 2} \bigvee_{2n \times 2n} \bigvee_{12 \times 2}$$

$$910 \times (A) = 11$$

$$00 \times (P) = 11$$

$$V = [0, ..., 0_{12}]$$

$$P_{2}$$

$$P_{3}$$

$$C = [0, T]$$

$$\frac{2\pi x^{2}}{2\pi x^{2}}$$

$$\frac{2\pi$$

Solution Watch lecture https://drive.google.com/file/d/1cY02DTagpckbY15gS0PYBu569vlZUNN6/view?usp=sharing

1. Write cross product as a matrix operation

$$[\underline{\mathbf{u}}_i]_{\times} = \begin{bmatrix} 0 & -w_i & y_i \\ w_i & 0 & -x_i \\ -y_i & x_i & 0 \end{bmatrix}$$

2. Write $P\underline{\mathbf{X}}_i$ in terms of row vectors.

$$P\underline{\mathbf{X}}_{i} = \begin{bmatrix} \mathbf{p}_{1}^{\top} \\ \mathbf{p}_{2}^{\top} \\ \mathbf{p}_{3}^{\top} \end{bmatrix} \underline{\mathbf{X}}_{i} = \begin{bmatrix} \mathbf{p}_{1}^{\top}\underline{\mathbf{X}}_{i} \\ \mathbf{p}_{2}^{\top}\underline{\mathbf{X}}_{i} \\ \mathbf{p}_{3}^{\top}\underline{\mathbf{X}}_{i} \end{bmatrix}$$

3. Note that all the three terms like $\mathbf{p}_1^{\top} \underline{\mathbf{X}}_i$ are scalars hence they are symmetric. Hence $\mathbf{p}_1^{\top} \underline{\mathbf{X}}_i = \underline{\mathbf{X}}_i^{\top} \mathbf{p}_1$.

$$P\underline{\mathbf{X}}_{i} = \begin{bmatrix} \underline{\mathbf{X}}_{i}^{\top} \mathbf{p}_{1} \\ \underline{\mathbf{X}}_{i}^{\top} \mathbf{p}_{2} \\ \underline{\mathbf{X}}_{i}^{\top} \mathbf{p}_{3} \end{bmatrix}$$

4. Substitute these values in the original equation $\underline{\mathbf{u}}_i \times P\underline{\mathbf{X}}_i = \mathbf{0}_{3\times 1}$.

$$\begin{bmatrix} 0 & -w_i & y_i \\ w_i & 0 & -x_i \\ -y_i & x_i & 0 \end{bmatrix} \begin{bmatrix} \underline{\mathbf{X}}_i^{\top} \mathbf{p}_1 \\ \underline{\mathbf{X}}_i^{\top} \mathbf{p}_2 \\ \underline{\mathbf{X}}_i^{\top} \mathbf{p}_3 \end{bmatrix} = \mathbf{0}_{3 \times 1}$$

5. Matrix multiply

$$\begin{bmatrix} 0 - w_i \underline{\mathbf{X}}_i^{\top} \mathbf{p}_2 + y_i \underline{\mathbf{X}}_i^{\top} \mathbf{p}_3 \\ w_i \underline{\mathbf{X}}_i^{\top} \mathbf{p}_1 + 0 - x_i \underline{\mathbf{X}}_i^{\top} \mathbf{p}_3 \\ -y_i \underline{\mathbf{X}}_i^{\top} \mathbf{p}_1 + x_i \underline{\mathbf{X}}_i^{\top} \mathbf{p}_2 + 0 \end{bmatrix} = \mathbf{0}_{3 \times 1}$$

6. Write the unknowns as a single vector, and the knowns as a matrix multiplication with the unknowns

$$\begin{bmatrix} \mathbf{0}^{\top} & -w_i \underline{\mathbf{X}}_i^{\top} & y_i \underline{\mathbf{X}}_i \\ w_i \underline{\mathbf{X}}_i^{\top} & \mathbf{0}^{\top} & -x_i \underline{\mathbf{X}}_i^{\top} \\ -y_i \underline{\mathbf{X}}_i^{\top} & x_i \underline{\mathbf{X}}_i^{\top} & \mathbf{0}^{\top} \end{bmatrix}_{3 \times 12} \begin{bmatrix} \mathbf{p}_1 \\ \mathbf{p}_2 \\ \mathbf{p}_3 \end{bmatrix}_{12 \times 1} = \mathbf{0}_{3 \times 1}$$

7. Pick only two of the equations as only two are linearly independent.

$$\begin{bmatrix} \mathbf{0}^{\top} & -w_i \underline{\mathbf{X}}_i^{\top} & y_i \underline{\mathbf{X}}_i \\ w_i \underline{\mathbf{X}}_i^{\top} & \mathbf{0}^{\top} & -x_i \underline{\mathbf{X}}_i^{\top} \end{bmatrix}_{2 \times 12} \begin{bmatrix} \mathbf{p}_1 \\ \mathbf{p}_2 \\ \mathbf{p}_3 \end{bmatrix}_{12 \times 1} = \mathbf{0}_{2 \times 1}$$

8. Collect all the equations from n pairs of corresponding points $\underline{\mathbf{u}}_1, \dots, \underline{\mathbf{u}}_n$ and $\underline{\mathbf{X}}_1, \dots, \underline{\mathbf{X}}_n$.

$$\begin{bmatrix} \mathbf{0}^{\top} & -w_1 \underline{\mathbf{X}}_1^{\top} & y_1 \underline{\mathbf{X}}_1 \\ w_1 \underline{\mathbf{X}}_1^{\top} & \mathbf{0}^{\top} & -x_1 \underline{\mathbf{X}}_1^{\top} \\ \vdots & \vdots & \vdots \\ \mathbf{0}^{\top} & -w_n \underline{\mathbf{X}}_n^{\top} & y_n \underline{\mathbf{X}}_n \\ w_n \underline{\mathbf{X}}_n^{\top} & \mathbf{0}^{\top} & -x_n \underline{\mathbf{X}}_n^{\top} \end{bmatrix}_{2n \times 12} \begin{bmatrix} \mathbf{p}_1 \\ \mathbf{p}_2 \\ \mathbf{p}_3 \end{bmatrix}_{12 \times 1} = \mathbf{0}_{2n \times 1}$$

3

9. P matrix has rank rank(P) = 11 because it has 12 elements and equivalence upto a scale factor. So the solution of the above equation can be computed from SVD by choosing the right singular vector corresponding to the smallest singular value.

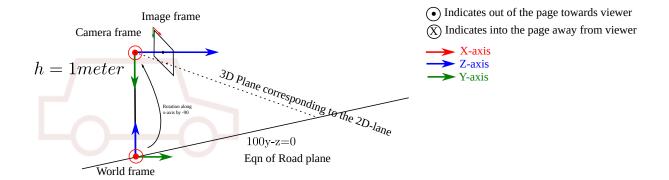
$$A = \begin{bmatrix} \mathbf{0}^{\top} & -w_1 \underline{\mathbf{X}}_1^{\top} & y_1 \underline{\mathbf{X}}_1 \\ w_1 \underline{\mathbf{X}}_1^{\top} & \mathbf{0}^{\top} & -x_1 \underline{\mathbf{X}}_1^{\top} \\ \vdots & \vdots & \vdots \\ \mathbf{0}^{\top} & -w_n \underline{\mathbf{X}}_n^{\top} & y_n \underline{\mathbf{X}}_n \\ w_n \underline{\mathbf{X}}_n^{\top} & \mathbf{0}^{\top} & -x_n \underline{\mathbf{X}}_n^{\top} \end{bmatrix} = U \Sigma V^T$$

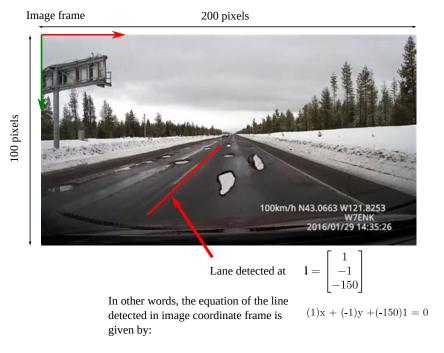
Let
$$V = [\mathbf{v}_1, \dots, \mathbf{v}_{\mathbf{v}}]$$
, then

$$egin{bmatrix} \mathbf{p}_1 \ \mathbf{p}_2 \ \mathbf{p}_3 \end{bmatrix} = \mathbf{v}_{\mathbf{p}}$$

Now we can write the P matrix as

$$P = \begin{bmatrix} \mathbf{p}_1^\top \\ \mathbf{p}_2^\top \\ \mathbf{p}_3^\top \end{bmatrix}$$





 $K = \begin{bmatrix} 100 & 0 & 100 \\ 0 & 100 & 50 \\ 0 & 0 & 1 \end{bmatrix}$

Figure 1: Line-plane triangulation

Problem 2 In figure 1 find the 3D representation of the lane the World coordinate frame, in terms of h (the height of the camera), image-representation of the line 1 (provided in figure), camera matrix K (provided in figure). Assume the lane to be a straight line. The Camera is mounted directly on top of the world frame, both of which are aligned to the gravity vector. The road is a perfect plane with a slope such that the equation of road plane in world-coordinate frame is given by $100Y_w - Z_w = 0$ and the lane lies on the road plane. Provide the formula or pseudo-code for computing the 3D representation of the lane, and also substitute in the values. (20 min, 20 marks)

Solution Watch lecture https://drive.google.com/file/d/1JaEwLxQ2BvT30sVxshmglBz_v5FjC1hW/view?usp=sharing See homework 4 solution.

.

$$f(u) = 2u^{T}A^{T}Au - 3v^{T}b + t^{T}Cu + d^{T}CR$$

$$\frac{\partial f(u)}{\partial u} = 0$$

$$\frac{\partial f(u)}{\partial u} =$$

$$\frac{\partial f(u)}{\partial u} = \frac{4A^{T}Au}{4A^{T}(3b-4)} = 0$$

Problem 3 Find the minimum point of the function, $f(\mathbf{u}) = 2\mathbf{u}^{\top}A^{\top}A\mathbf{u} - 3\mathbf{u}^{\top}\mathbf{b} + 4\mathbf{c}^{\top}\mathbf{u} + d$. Let $\mathbf{u} \in \mathbb{R}^{n \times 1}$ be a n-dimensional vector and sizes of $A, \mathbf{b}, \mathbf{c}, d$ be such that matrix multiplication and addition is valid. Also assume that $A^{\top}A$ is full rank, hence invertible.

Solution Watch this lecture https://drive.google.com/file/d/1wgY2LAw7LQnh_IyHY0XDAr2yorXta93Z/view?usp=sharing

Problem 4 Let matrix $A \in \mathbb{R}^{m \times n}$ be a $m \times n$ matrix. We are given that $B = A^{\top}A$ has n orthonormal eigen vectors $\mathbf{e}_1, \dots \mathbf{e}_n$ with corresponding eigen values as $\lambda_1 \dots \lambda_n$ such that $B\mathbf{e}_i = \lambda_i \mathbf{e}_i$ for all $i \in \{1, \dots n\}$. Let the rank of matrix A be r. Write the thin singular value decomposition of $A = U_{m \times r} \Sigma_{r \times r} V_{n \times r}^{\top}$ in terms of eigen values and eigen vectors of matrix $B = A^{\top}A$.

Solution Go through these slides. Watch this lecture https://drive.google.com/file/d/13a0_XI7kykQNOs5RJOfhqtL5fview?usp=sharing

The matrix of right singular vectors of A is same as the eigen vector matrix of $B = A^{T}A$.

$$V = \left[\mathbf{e}_1 \dots \mathbf{e}_r \right] \in \mathbb{R}^{n \times r} \tag{1}$$

The matrix of singular values are the square root of eigen values of B.

$$\Sigma = \begin{bmatrix} \sqrt{\lambda_1} & \dots & 0 \\ 0 & \ddots & 0 \\ 0 & \dots & \sqrt{\lambda_r} \end{bmatrix} \in bbR^{r \times r}$$
 (2)

$$U = \left[\mathbf{u}_1 \dots \mathbf{u}_r\right] \in \mathbb{R}^{m \times r} \tag{3}$$

where
$$\mathbf{u}_i = \frac{A\mathbf{e}_i}{\sqrt{\lambda_i}}$$
 (4)

A=
$$U \ge V^T$$

$$V = E$$

$$V = \begin{bmatrix} e_1 & \cdots & e_n \end{bmatrix}$$

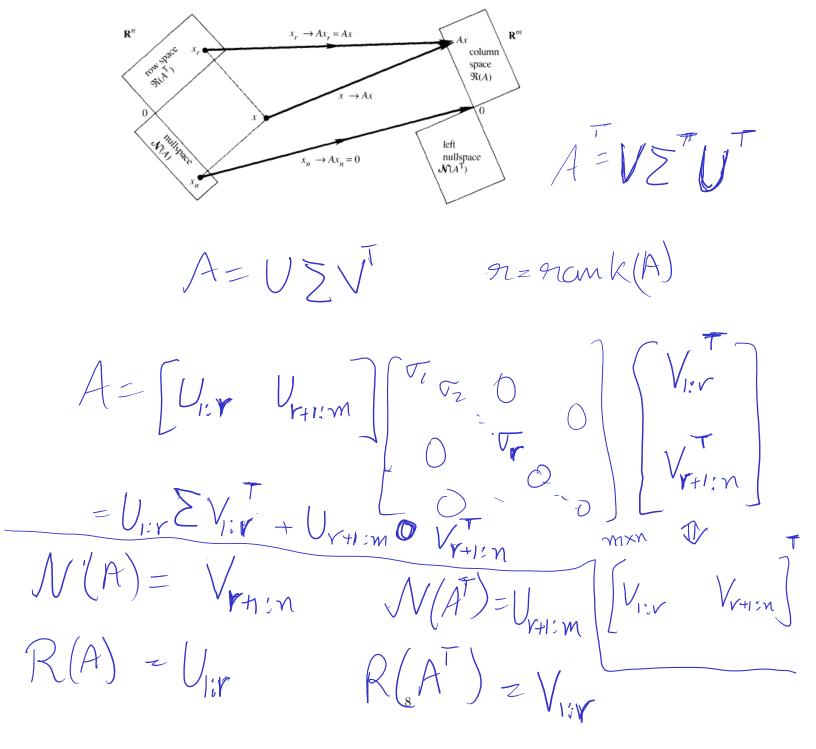
V= [U, -- Um] mxm

defn

Problem 5 Let matrix $A \in \mathbb{R}^{m \times n}$ has the singular value decomposition (SVD) as $A = U\Sigma V^{\top}$ and rank of the matrix be r = rank(A). Write the basis vectors of the four fundamental subspaces of matrix A in terms of SVD,

- 1. Null space of A ($\mathcal{N}(A) = ?$).
- 2. Column space or range space $(\mathcal{R}(A) =?)$.
- 3. Row space $(\mathcal{R}(A^{\top}) = ?)$.
- 4. Left null space $(\mathcal{N}(A^{\top}) = ?)$.

You can denote the first r column vectors of U has $U_{1:r} \in \mathbb{R}^{m \times r}$ and the renaming m-r vectors as $U_{r+1:m} \in \mathbb{R}^{m \times (m-r)}$. Similarly for V, first r column vectors of $V_{1:r} \in \mathbb{R}^{n \times r}$ and $V_{r+1:n} \in \mathbb{R}^{n \times (n-r)}$.



Solution Watch lecture https://drive.google.com/file/d/17Cr0rMr567gfRNrrj9ri9vtsFaNLqW7t/view?usp=sharing

- 1. Null space of A ($\mathcal{N}(A) = V_{r+1:n}$).
- 2. Column space or range space $(\mathcal{R}(A) = U_{1:r})$.
- 3. Row space $(\mathcal{R}(A^{\top}) = V_{1:r})$.
- 4. Left null space $(\mathcal{N}(A^{\top}) = U_{r+1:m})$.

Extra practice problems

Problem 6 Find a line passing through the following points

$$\mathbf{u}_1 = [101, 203]^{\top}, \mathbf{u}_2 = [49, 102]^{\top}, \mathbf{u}_3 = [27, 51]^{\top}, \mathbf{u}_4 = [201, 403]^{\top}, \mathbf{u}_5 = [74, 151]^{\top}.$$

You can leave the output in terms of SVD.

Problem 7 Find a plane passing through the following points

$$\mathbf{x}_1 = [9.99, 101, 203]^{\top}, \mathbf{x}_2 = [5.1, 49, 102]^{\top}, \mathbf{x}_3 = [2.5, 27, 51]^{\top}, \mathbf{x}_4 = [21, 201, 403]^{\top}, \mathbf{x}_5 = [7.6, 74, 151]^{\top}.$$

You can leave the output in terms of SVD.

Problem 8 Find the 3D line in parameteric representation that is formed by the intersection of two planes $\mathbf{p}^{\top}\underline{\mathbf{x}} = 0$ (with $\mathbf{p} = [1, 2, 3, 4]^{\top}$) and $\mathbf{q}^{\top}\underline{\mathbf{x}} = 0$ where $\mathbf{q} = [-3, 2, 1, 4]^{\top}$.

Problem 9 Find the point on the intersection of following 3D lines $\mathbf{x} = \lambda_1 \mathbf{d}_1 + \mathbf{y}$ and $\mathbf{x} = \lambda_2 \mathbf{d}_2 + \mathbf{z}$. Here $\lambda_1 \in \mathbb{R}$ and $\lambda_2 \in \mathbb{R}$ are the free parameters. The rest of the parameters have the following values

$$\mathbf{d}_1 = [1, 2, 0]^{\top}, \mathbf{d}_2 = [-2, 1, 0]^{\top}, \mathbf{y} = [1, 2, 0]^{\top}, \mathbf{z} = [4, 5, 0]^{\top}$$

Problem 10 Find the point of intersection of the 3D line $\mathbf{x} = \lambda \mathbf{d} + \mathbf{x}_0$ with the 3D plane $\mathbf{p}^{\top} \underline{\mathbf{x}} = 0$. The parameters have the following

$$\mathbf{d} = [1, 2, 0]^\top, \mathbf{x}_0 = [3, 4, 5]^\top, \mathbf{p} = [1, 2, 0, 7]^\top$$

$$\mathcal{X}_{1} = \begin{bmatrix} 9.99 \\ 10B \\ 203 \end{bmatrix} \quad \mathcal{X}_{2} = \begin{bmatrix} 5.1 \\ 49 \\ 102 \end{bmatrix}, \quad -1.$$

$$\frac{\chi_{1}}{\chi_{2}} = 0$$

$$\frac{\chi_{2}}{\chi_{5}} = 0$$

$$\frac{\chi_{5}}{\chi_{5}} = 0$$

$$\frac{\chi_{5}}{\chi_{5}} = 0$$

$$\frac{\chi_{5}}{\chi_{5}} = 0$$

$$A = UZV_{4x4}^T$$

pz V4

Solm &

$$\chi = \frac{y_3}{\sqrt{3}} + \frac{y_4}{\sqrt{3}} = \frac{y_3}{\sqrt{3}} + \frac{t}{\sqrt{2}} \frac{y_4}{\sqrt{3}}$$

HWH + HW5 way

$$\begin{pmatrix}
P_{1:3} \\
P_{1:3}
\end{pmatrix}$$

$$2C = \lambda \begin{pmatrix}
P_{1:3} \\
P_{1:3}
\\
P_{1:3}
\end{pmatrix}$$

$$+ \begin{pmatrix}
P_{1:3} \\
P_{1:3}
\\
P_{1:3}
\end{pmatrix}$$

$$-P_{4}$$

$$A \chi = 0$$

Solm?
$$z = \lambda_1 d_1 + y$$

$$z = \lambda_2 d_2 + z$$

$$\lambda_1 d_1 + y = \lambda_2 d_2 + z$$

$$\lambda_1 d_1 + y = \lambda_2 d_2 + z$$

$$\lambda_1 d_1 - d_2 = z - y$$

$$\left[\frac{d_1}{d_1} - \frac{d_2}{d_2} \right] \left[\frac{\lambda_1}{\lambda_2} \right] z = z - y$$

$$A = \left[\frac{A^T A}{\lambda_1} \right]^T A^T (z - y)$$

$$A = \left[\frac{A^T A}{\lambda_2} \right]^T A^T (z - y)$$

$$2c = \left[\frac{d_1 + y}{\lambda_1} \right] + \left(\frac{d_2 d_2 + z}{\lambda_2} \right) \left[\frac{d_1 + y}{\lambda_2} \right]$$

$$2c = \left[\frac{d_1 + y}{\lambda_1} \right] + \left(\frac{d_2 d_2 + z}{\lambda_2} \right) \left[\frac{d_1 + y}{\lambda_2} \right]$$

元 マンカースの Solm 10 -30 lmi px = 0_30 plum Pr3 7 + Py = 0 PT () d + 20) + py = 0 > (P13 d)+ P13 76+ Py=0 $\lambda = \begin{bmatrix} -P_{1:3} & -P_{4} \\ P_{1:3} & 0 \end{bmatrix}$ 元マンオーな

Practice problem solutions

Solution 6 Watch lecture https://drive.google.com/file/d/13a0_XI7kykQNOs5RJOfhqtL5fgEzVfkN/view?usp=sharing Let $l \in \mathbb{P}^2$ be the parameters of the line, so that $\mathbf{u}^{\top} \mathbf{l} = 0$.

$$A = \begin{bmatrix} \mathbf{u}_1^\top & 1 \\ \mathbf{u}_2^\top & 1 \\ \mathbf{u}_3^\top & 1 \\ \mathbf{u}_4^\top & 1 \\ \mathbf{u}_5^\top & 1 \end{bmatrix}_{5\times 3}$$

We are looking for the solution of $A\mathbf{l} = 0$. Let the SVD of $A = U\Sigma V^{\top}$. Let $V = [\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3]$, then the representation of the line $\mathbf{l} = \mathbf{v}_3$.

Solution 7 Watch the lecture https://drive.google.com/file/d/1PEdgHCf6UdOWYMVtCRlsjfctDcnBm8E9/view?usp=sharing

Let $\mathbf{p} \in \mathbb{P}^3$ be the parameters of the plane, so that $\mathbf{\underline{x}}^{\mathsf{T}}\mathbf{p} = 0$.

$$A = \begin{bmatrix} \mathbf{x}_1^\top & 1 \\ \mathbf{x}_2^\top & 1 \\ \mathbf{x}_3^\top & 1 \\ \mathbf{x}_4^\top & 1 \\ \mathbf{x}_5^\top & 1 \end{bmatrix}_{5 \times 4}$$

We are looking for the solution of $A\mathbf{p} = 0$. Let the SVD of $A = U\Sigma V^{\top}$. Let $V = [\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3, \mathbf{v}_4]$, then the representation of the line $\mathbf{p} = \mathbf{v}_4$.

 $\begin{aligned} & \textbf{Solution 8} & \textbf{Watch the lecture https://drive.google.com/file/d/1JaEwLxQ2BvT30sVxshmg1Bz_v5FjC1hW/view?usp=sharing } \mathbf{x} = \lambda(\mathbf{p}_{1:3} \times \mathbf{q}_{1:3}) + \begin{bmatrix} \mathbf{p}_{1:3}^{\top} \\ \mathbf{q}_{1:3} \end{bmatrix}^{\dagger} \begin{bmatrix} -p_4 \\ q_4 \end{bmatrix} \end{aligned}$

 $\textbf{Solution 9} \quad \text{Watch the lecture https://drive.google.com/file/d/1foVVQBC0krljjJ3f-UP4zsHn4l-c611e/view?usp=sharing }$

$$egin{bmatrix} \left[\mathbf{d}_1 & -\mathbf{d}_2
ight] egin{bmatrix} \lambda_1 \ \lambda_2 \end{bmatrix} = \mathbf{z} - \mathbf{y}$$

or

$$egin{bmatrix} \lambda_1 \ \lambda_2 \end{bmatrix} = egin{bmatrix} \mathbf{d}_1 & -\mathbf{d}_2 \end{bmatrix}^\dagger \mathbf{z} - \mathbf{y}$$

The point of intersection is given by

$$\mathbf{x} = \lambda_1 \mathbf{d}_1 + \mathbf{y}$$

 $Solution \ 10 \quad \text{Watch the lecture https://drive.google.com/file/d/1wgY2LAw7LQnh_IyHY0XDAr2yorXta93Z/view?usp=sharing}$

$$\lambda \mathbf{p}_{1:3}^{\top} \mathbf{d} + \mathbf{p}_{1:3}^{\top} \mathbf{x}_0 + p_4 = 0$$

Solve for λ .

$$\lambda = -\frac{\mathbf{p}_{1:3}^{\top} \mathbf{x}_0 + p_4}{\mathbf{p}_{1:3}^{\top} \mathbf{d}}$$

Point of intersection is

$$\mathbf{x} = \lambda \mathbf{d} + \mathbf{x}_0$$