

ECE 417/598: Null space, Singular Value Decomposition

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Homogeneous representation of lines

$$\mathbb{P}^2 = \mathbb{R}^3 - \{(0, 0, 0)^\top\}$$

$$ax + by + 1.c = 0$$

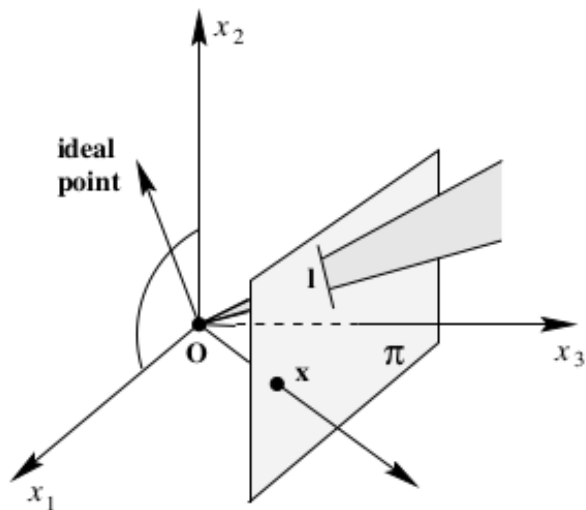
$$\mathbf{l} = \begin{bmatrix} a \\ b \\ c \end{bmatrix}$$

$$\mathbf{x} = \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

The point $\mathbf{x} \in \mathbb{P}^2$ lies on a line \mathbf{l} if and only if

$$\mathbf{l}^\top \mathbf{x} = 0$$

Points are rays and lines are planes



Intersection of lines

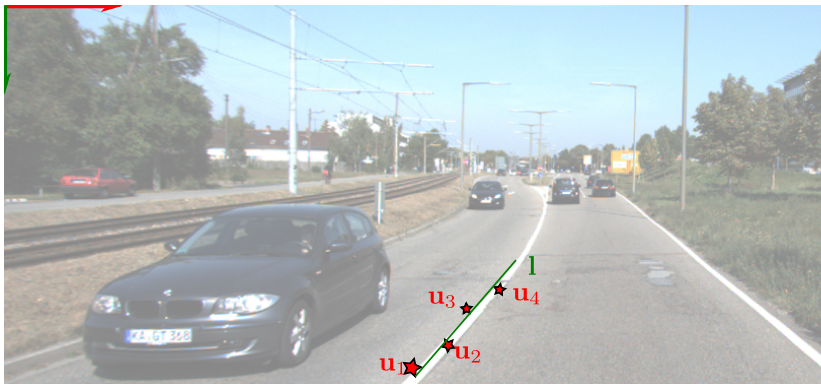
Two line \mathbf{l}_1 and \mathbf{l}_2 intersect at $\mathbf{x} \in \mathbb{P}^2$

$$\mathbf{x} = \mathbf{l}_1 \times \mathbf{l}_2$$

Line joining points

Two point \mathbf{x}_1 and \mathbf{x}_2 form a $\mathbf{l} \in \mathbb{P}^2$

$$\mathbf{l} = \mathbf{x}_1 \times \mathbf{x}_2$$



$$\underline{\mathbf{u}}_1 = [100, 98, 1]^\top$$

$$\underline{\mathbf{u}}_2 = [105, 95, 1]^\top$$

$$\underline{\mathbf{u}}_3 = [107, 90, 1]^\top$$

$$\underline{\mathbf{u}}_4 = [110, 85, 1]^\top$$

Find the line \mathbf{l} such that it is the “closest line” passing through $\underline{\mathbf{u}}_1, \dots, \underline{\mathbf{u}}_4$.

$$U = \begin{bmatrix} \mathbf{u}_1^\top \\ \mathbf{u}_2^\top \\ \mathbf{u}_3^\top \\ \mathbf{u}_4^\top \end{bmatrix}$$

We want to solve for \mathbf{l} such that

$$U\mathbf{l} = \mathbf{0}$$

The column space (also called the range) of matrix $A \in \mathbb{R}^{m \times n}$, denoted by $\mathcal{R}(A)$ is defined as the set of all vectors $\mathbf{b} \in \mathbb{R}^m$ that can be generated by $\mathbf{b} = A\mathbf{x}$ where $\mathbf{x} \in \mathbb{R}^n$, that is,

$$\mathcal{R}(A) = \{\mathbf{b} \mid \mathbf{b} = A\mathbf{x} \text{ for all } \mathbf{x} \in \mathbb{R}^n\}. \quad (1)$$

The nullspace of $A \in \mathbb{R}^{m \times n}$ is defined as the set of all vectors $\mathbf{x} \in \mathbb{R}^n$ such that $A\mathbf{x} = \mathbf{0}_m$. In other words,

$$\mathcal{N}(A) = \{\mathbf{x} \in \mathbb{R}^n \mid \mathbf{0}_m = A\mathbf{x}\}. \quad (2)$$

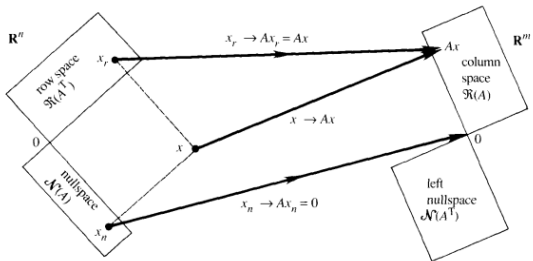
The task of finding the column space or the null space is the task of finding the minimal set of vectors that *span* the vector spaces $\mathcal{R}(A)$ or $\mathcal{N}(A)$ respectively.

Find the $\mathcal{R}(A)$ and $\mathcal{N}(A)$ of the matrix A

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 2 & 4 & 6 \end{bmatrix} = \begin{bmatrix} \mathbf{r}_1^\top \\ \mathbf{r}_2^\top \\ \mathbf{r}_3^\top \end{bmatrix}$$

Four fundamental subspaces of matrix $A \in \mathbb{R}^{m \times n}$:

1. Column space: All possible values of $\mathbf{b} = A\mathbf{x}$ for any $\mathbf{x} \in \mathbb{R}^n$.
2. Null space: All possible values of $\mathbf{x} \in \mathbb{R}^n$ so that $A\mathbf{x} = \mathbf{0}_m$.
3. Row space: Column space of A^\top . All possible values of $\mathbf{b} = A^\top \mathbf{x}$ for any $\mathbf{x} \in \mathbb{R}^m$.
4. Left Null space: Null space of A^\top . All possible values of $\mathbf{y} \in \mathbb{R}^m$ so that $\mathbf{y}^\top \mathbf{A} = 0$.



The four fundamental subspaces of A

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 2 & 4 & 6 \end{bmatrix}$$

Geometric intuition

Eigenvalues and Eigenvectors

For a square matrix A , the λ_i and \mathbf{x}_i that satisfy the following equation are called eigenvalues and eigenvectors respectively.

$$A\mathbf{x} = \lambda\mathbf{x} \text{ or } (A - \lambda I)\mathbf{x} = 0 \quad (3)$$

λ is chosen to ensure that $A - \lambda I$ has null space, hence, characteristic equation

$$\det(A - \lambda I) = 0 \quad (4)$$

For symmetrix matrix $A = A^\top$, eigenvalues are real, and eigenvectors are orthonormal,

$$A[\mathbf{x}_1, \dots, \mathbf{x}_n] = [\mathbf{x}_1, \dots, \mathbf{x}_n] \begin{bmatrix} \lambda_1 & \dots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \dots & \lambda_n \end{bmatrix} \quad (5)$$

$$AS = SA \quad (6)$$

$$\text{if } A = A^\top \text{ then } A = S\Lambda S^\top \quad (7)$$

Numerical example

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 2 & 4 & 6 \end{bmatrix}$$

Find eigen values and eigen vectors.

Singular Value Decomposition (SVD)

$$A = U \begin{bmatrix} \Sigma & 0 \\ 0 & 0 \end{bmatrix} V^{\top} \quad (8)$$

$$A^{\top} A = V \Sigma^2 V^{-1} \quad (9)$$

$$A^{\top} A \mathbf{v}_i = \lambda_i \mathbf{v}_i \quad \lambda_i = \sigma_i^2 \quad (10)$$

$$AV = U \begin{bmatrix} \Sigma & 0 \\ 0 & 0 \end{bmatrix} \quad (11)$$

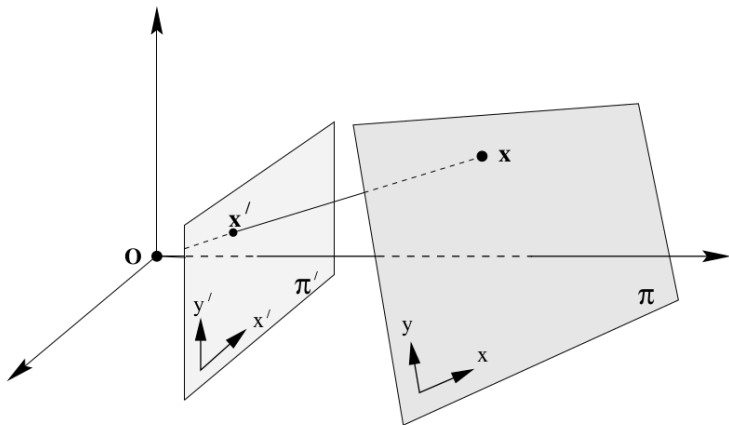
$$U^+ = \Sigma^{-1} AV^+ \quad (12)$$

Numerical example

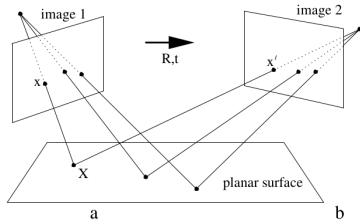
Find singular value decomposition

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 2 & 4 & 6 \end{bmatrix}$$

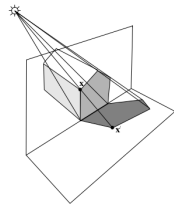
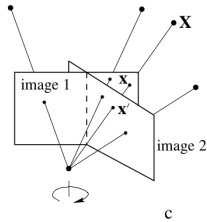
Homography



Examples of Homography



b





Computing Homography



Computing Homography



Solving for Homography derivation

Direct Linear Transformation (DLT) algorithm

Objective

Given $n \geq 4$ 2D to 2D point correspondences $\{\mathbf{x}_i \leftrightarrow \mathbf{x}'_i\}$, determine the 2D homography matrix \mathbf{H} such that $\mathbf{x}'_i = \mathbf{H}\mathbf{x}_i$.

Algorithm

- (i) For each correspondence $\mathbf{x}_i \leftrightarrow \mathbf{x}'_i$ compute the matrix \mathbf{A}_i from (4.1). Only the first two rows need be used in general.
- (ii) Assemble the $n \times 2 \times 9$ matrices \mathbf{A}_i into a single $2n \times 9$ matrix \mathbf{A} .
- (iii) Obtain the SVD of \mathbf{A} (section A4.4(p585)). The unit singular vector corresponding to the smallest singular value is the solution \mathbf{h} . Specifically, if $\mathbf{A} = \mathbf{U}\mathbf{D}\mathbf{V}^T$ with \mathbf{D} diagonal with positive diagonal entries, arranged in descending order down the diagonal, then \mathbf{h} is the last column of \mathbf{V} .
- (iv) The matrix \mathbf{H} is determined from \mathbf{h} as in (4.2).

2D homography

Given a set of points $\mathbf{x}_i \in \mathbb{P}^2$ and a corresponding set of points $\mathbf{x}'_i \in \mathbb{P}^2$, compute the projective transformation that takes each \mathbf{x}_i to \mathbf{x}'_i . In a practical situation, the points \mathbf{x}_i and \mathbf{x}'_i are points in two images (or the same image), each image being considered as a projective plane \mathbb{P}^2 .

3D to 2D camera projection matrix estimation

Given a set of points \mathbf{X}_i in 3D space, and a set of corresponding points \mathbf{x}_i in an image, find the 3D to 2D projective \mathbf{P} mapping that maps \mathbf{X}_i to $\mathbf{x}_i = \mathbf{P}\mathbf{X}_i$.