

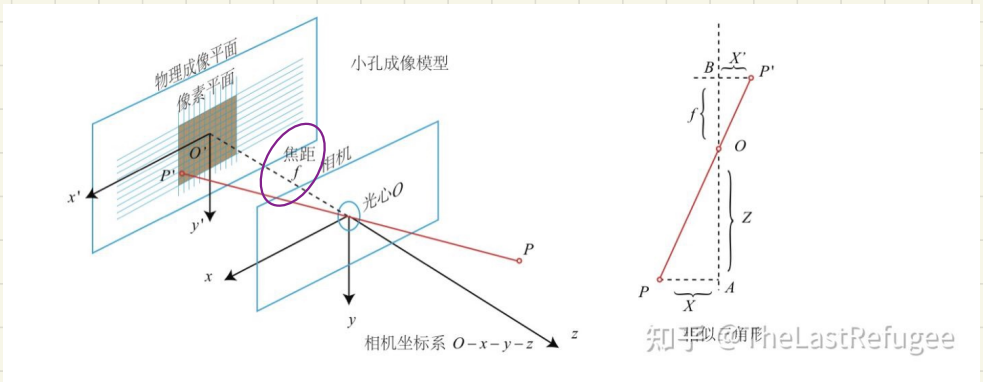
# Intrinsic matrix and Extrinsic matrix

World  $\xleftrightarrow{\text{Ext}}$  Camera cord  $\xleftrightarrow{\text{Int}}$  Pixel cord

$$S \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} \alpha_x & 0 & u_0 \\ 0 & \alpha_y & v_0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix} \begin{bmatrix} t_1 \\ t_2 \\ t_3 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

$$= \underbrace{K}_{\text{Int}} \underbrace{[R \ t]}_{\text{Ext}} \underbrace{Q}_{\text{世界坐标系下的某一点}}$$

Int :



$\therefore$  similar triangle

$$\therefore \frac{z}{f} = \frac{x}{x'} = \frac{y}{y'}$$

Suppose a point  $P = [X, Y, Z]^T$  in the world cord

and a  $P' = [X', Y', Z']^T$  in the pixel cord.

So, we can get

$$X' = X \frac{f}{z}, \quad Y' = Y \frac{f}{z}$$

The above functions describe the relation between the world cord and pixel cord.

But the original point of pixel cord is at the upper left corner,

So, we suppose  $P'$  cord is  $[u, v]^T$  in the pixel cord, and we can get

$$u = \alpha X' + C_x$$

$\alpha, B$  表示縮放係數

$$v = \beta Y' + C_y$$

$C_x, C_y$  表示平移

$$u = \alpha X \frac{f}{z} + C_x = X \frac{f_x}{z} + C_x$$

$$v = \beta Y \frac{f}{z} + C_y = Y \frac{f_y}{z} + C_y$$

$f_x$  and  $f_y$  unit is pixel

We can use homogeneous coord rewritten the formula.

$$\begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \frac{1}{Z} \begin{bmatrix} f_x & 0 & c_x \\ 0 & f_y & c_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \end{bmatrix}$$

Ext:

Int matrix only tells us how to transform between camera coord and pixel coord. We still need a matrix to transform between camera coord and world coord.

$$P = \underbrace{R}_{\text{旋轉矩陣}} \underbrace{P_w}_{\text{世界座標系下的點}} + \underbrace{t}_{\text{平移矩陣}}$$

$$= \begin{bmatrix} r_{11} & r_{12} & r_{13} & t_1 \\ r_{21} & r_{22} & r_{23} & t_2 \\ r_{31} & r_{32} & r_{33} & t_3 \end{bmatrix} \begin{bmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{bmatrix}$$

World coord to Camera coord

∴

$$P_{uw} = \frac{1}{Z} \begin{bmatrix} f_x & 0 & c_x \\ 0 & f_y & c_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} r_{11} & r_{12} & r_{13} & t_1 \\ r_{21} & r_{22} & r_{23} & t_2 \\ r_{31} & r_{32} & r_{33} & t_3 \end{bmatrix} \begin{bmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{bmatrix}$$

繞 X 軸:

$$R_x(\theta_x) = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos\theta_x & -\sin\theta_x \\ 0 & \sin\theta_x & \cos\theta_x \end{bmatrix}$$

繞 Y 軸

$$R_y(\theta_y) = \begin{bmatrix} \cos\theta_y & 0 & \sin\theta_y \\ 0 & 1 & 0 \\ -\sin\theta_y & 0 & \cos\theta_y \end{bmatrix}$$

繞 Z 軸

$$R_z(\theta_z) = \begin{bmatrix} \cos\theta_z & -\sin\theta_z & 0 \\ \sin\theta_z & \cos\theta_z & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$R = R_x(\theta_x) R_y(\theta_y) R_z(\theta_z)$$

$P' = M P_w = K [R \ T] P_w$

5個未知數 (pointing to K)  
 3個未知數  $\theta_x, \theta_y, \theta_z$  (pointing to R)  
 3個未知數  $t_x, t_y, t_z$  (pointing to T)

$3 \times 4$  (pointing to M)  
 $4 \times 1$  (pointing to  $P_w$ )

內參 (pointing to R)  
 外參 (pointing to T)

$3 \times 4$  (pointing to M)  
 $4 \times 4$  (pointing to  $[R \ T]$ )  
 $4 \times 1$  (pointing to  $P_w$ )

$5 + 3 + 3 = 11$   
 總共有 11 個未知數

$$M = \begin{bmatrix} m_{11} & m_{12} & m_{13} & m_{14} \\ m_{21} & m_{22} & m_{23} & m_{24} \\ m_{31} & m_{32} & m_{33} & m_{34} \end{bmatrix} = \begin{bmatrix} m_1 \\ m_2 \\ m_3 \end{bmatrix}$$

$1 \times 4$  (pointing to  $m_1$ )  
 $1 \times 4$  (pointing to  $m_2$ )  
 $1 \times 4$  (pointing to  $m_3$ )

投影矩陣

$$P' = \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} m_{11} & m_{12} & m_{13} & m_{14} \\ m_{21} & m_{22} & m_{23} & m_{24} \\ m_{31} & m_{32} & m_{33} & m_{34} \end{bmatrix} \begin{bmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{bmatrix}$$

$P_i$   
 $4 \times 1$  (pointing to the vector  $\begin{bmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{bmatrix}$ )

$$= \begin{bmatrix} m_{11} X_w + m_{12} Y_w + m_{13} Z_w + m_{14} \\ m_{21} X_w + m_{22} Y_w + m_{23} Z_w + m_{24} \\ m_{31} X_w + m_{32} Y_w + m_{33} Z_w + m_{34} \end{bmatrix}$$

$$= \begin{bmatrix} m_1 P_i \\ m_2 P_i \\ m_3 P_i \end{bmatrix} \xrightarrow{H \rightarrow E} \begin{bmatrix} m_1 P_i \\ m_3 P_i \\ m_2 P_i \\ m_3 P_i \end{bmatrix}$$

$$M = \begin{bmatrix} m_{11} & m_{12} & m_{13} & m_{14} \\ m_{21} & m_{22} & m_{23} & m_{24} \\ m_{31} & m_{32} & m_{33} & m_{34} \end{bmatrix}$$

共有 11 個未知數

3x4

而一對點  $\begin{bmatrix} u_i \\ v_i \end{bmatrix} = \begin{bmatrix} \frac{m_{1i} p_i}{m_{3i} p_i} \\ \frac{m_{2i} p_i}{m_{3i} p_i} \end{bmatrix}$  可以提供 2 個

方程，所以至少需要 6 個點

$$u_i = \frac{m_{1i} p_i}{m_{3i} p_i}, \quad u_i m_{3i} p_i = m_{1i} p_i, \quad u_i m_{3i} p_i - m_{1i} p_i = 0$$

$$v_i = \frac{m_{2i} p_i}{m_{3i} p_i}, \quad v_i m_{3i} p_i = m_{2i} p_i, \quad v_i m_{3i} p_i - m_{2i} p_i = 0$$

若有  $n$  個點，則：

$$\begin{bmatrix} u_1 m_{31} p_1 - m_{11} p_1 = 0 \\ v_1 m_{31} p_1 - m_{21} p_1 = 0 \\ u_2 m_{32} p_2 - m_{12} p_2 = 0 \\ v_2 m_{32} p_2 - m_{22} p_2 = 0 \\ \vdots \\ u_n m_{3n} p_n - m_{1n} p_n = 0 \\ v_n m_{3n} p_n - m_{2n} p_n = 0 \end{bmatrix}$$

$$P \stackrel{\text{def}}{=} \begin{bmatrix} p_1^T & 0^T & -u_1 p_1^T \\ 0^T & p_1^T & -v_1 p_1^T \\ \vdots & \vdots & \vdots \\ p_n^T & 0^T & -u_n p_n^T \\ 0^T & p_n^T & -v_n p_n^T \end{bmatrix}$$

2n x 12

$$M = \begin{bmatrix} m_{11}^T \\ m_{21}^T \\ m_{31}^T \\ \vdots \\ m_{1n}^T \\ m_{2n}^T \\ m_{3n}^T \end{bmatrix}$$

12 x 1

$$PM = \begin{bmatrix} P_1^T & 0^T & -u_1 P_1^T \\ \vdots & \vdots & \vdots \\ \vdots & \vdots & \vdots \end{bmatrix} \begin{bmatrix} m_1^T \\ m_2^T \\ m_3^T \end{bmatrix}$$

$$= \begin{bmatrix} \overset{1 \times 4}{\downarrow} P_1^T & \overset{4 \times 1}{\downarrow} m_1^T & \overset{1 \times 4}{\downarrow} -u_1 P_1^T & \overset{4 \times 1}{\downarrow} m_3^T \\ \vdots & \vdots & \vdots & \vdots \end{bmatrix}$$

已知  $PM = 0$  齊性線性方程

但不要  $0$  解，所以只能求非  $0$  解。

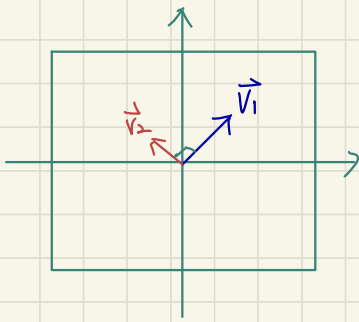
即 minimize  $\|PM\|$

SVD:

<https://m.youtube.com/watch?v=Ole48iAqh8E>

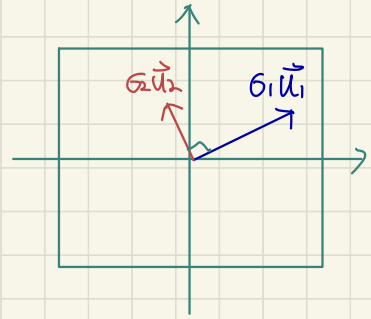
$$M = \underbrace{U}_{\text{旋轉}} \underbrace{\Sigma}_{\text{拉伸}} \underbrace{V^T}_{\text{旋轉}}$$

$M$  為一個矩陣，所以可以將其看成 linear transformation. 假設  $M$  為一  $2 \times 2$  矩陣.



$$V = [\vec{v}_1, \vec{v}_2]$$

$M$ -線性  
變換



$$U = [\vec{u}_1, \vec{u}_2]$$

$$\Sigma = \begin{bmatrix} \sigma_1 & 0 \\ 0 & \sigma_2 \end{bmatrix}$$

拉伸  
作用

$$U\Sigma = [\vec{u}_1\sigma_1, \vec{u}_2\sigma_2]$$

$$MV = U\Sigma$$

$$M = U\Sigma \underbrace{V^{-1}}_{\text{orthogonal matrix}}$$

$$= U\Sigma V^T$$



solve SVD:  $\neq (V)$

$$M = U \Sigma V^T$$

$$M^T M = (U \Sigma V^T)^T U \Sigma V^T$$
$$= (V^T)^T \Sigma^T U^T U \Sigma V^T$$

$$= \underbrace{V \Sigma^T U^T U}_{\Sigma} \Sigma V^T$$

$$(AB)^T = B^T A^T$$

$U$  是一個正交基底所以  
 $U^T = U^{-1}$ ,  $U^T U = I$

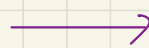
$\therefore$  是一個對稱矩陣  $\Sigma^T = \Sigma$

$$= V \Sigma \Sigma V^T$$

Let  $L = \Sigma \Sigma$ ,  $\Sigma = \begin{bmatrix} \sigma_1 & 0 \\ 0 & \sigma_2 \end{bmatrix}$

$$M^T M = V L V^T$$

$$\underline{M^T M (V)} = \underline{(V) L}$$



$M^T M$  特征向量公式

$$M^T M \vec{v} = \lambda \vec{v}$$

所以說,  $V$  就是  $M^T M$  的 eigen vector,

$L$  為 eigen value.

求  $V$ :

$$M = U \Sigma V^T$$

$$M M^T = U \Sigma V^T (U \Sigma V^T)^T$$

$$= U \Sigma V^T V \Sigma U^T$$

$$= U \Sigma \Sigma^T U^T$$

$$\text{let } \Sigma \Sigma = B$$

$$MM^T = U B U^T$$

$$MM^T U = U B$$

所以說,  $U$  就是  $MM^T$  的 eigen vector,  
 $B$  為 eigen value.

利用 SVD 求解  $\|PM\| = 0$ :

$$\min \|Pm\|^2 = \min \|U D V^T M\|^2 = \min \|D V^T M\|^2$$

└ 正交矩陣, 不影響 norm

約束條件  $\|x\| = 1$ :

將  $P$  進行 SVD 分解後, 由於奇異值越往右越小, 可將其視為 0. 而當奇異值中最小的數值為 0 的話, 則代表  $Pm=0$  有無限多組解。

proof:

$\therefore P = UDV^T$ ,  $D$  為  $PP^T$  or  $P^TP$  中的 eigen value, 且若矩陣的 eigen value 有 0 出現的話, 則可證明  $\text{rank}(A) < n$ , 具有無限多組解。

$$A = \begin{bmatrix} 2 & 1 \\ 2 & 1 \end{bmatrix}, \quad AX = 0, \quad \text{求 } X = ?$$

Eigen value of  $A$ :

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

$$\Rightarrow (2 - \lambda)(1 - \lambda) - 2 = 0$$

$$\Rightarrow 2 - 2\lambda - \lambda + \lambda^2 - 2 = 0$$

$$\Rightarrow \lambda^2 - 3\lambda = 0$$

$$\Rightarrow \lambda(\lambda - 3) = 0, \quad \lambda = \boxed{0}, 3$$

求  $X$ :

$$2x_1 + x_2 = 0, \quad x_1 = -\frac{1}{2}x_2$$

$$\text{null}(A) = \left\{ \begin{bmatrix} -\frac{1}{2} \\ 1 \end{bmatrix} \right\}$$

由上可知, 這樣會有無限多組解, 所以在求解時, 會為了方便計算, 求解, 設定一約束條件  $\|x\|=1$

證明  $P$  矩陣的最小奇異值對應的 Eigen vector 為  $\min \|Ax\|$  的最優解:

$$\min \|Ax\|^2 = \min \left\| \begin{matrix} r \times r & r \times r & r \times n & r \times 1 \\ U & D & V^T & x \end{matrix} \right\|^2 = \min \left\| \begin{matrix} r \times r & r \times n & r \times 1 \\ D & V^T & x \end{matrix} \right\|^2$$

└─ 正交矩陣, 不影響 norm

$$\text{令 } \begin{matrix} r \times 1 \\ r \times n & r \times 1 \end{matrix} y = \begin{matrix} r \times n & r \times 1 \\ V^T & x \end{matrix}, \quad \because \|x\|=1 \quad \therefore \|y\|=1$$

$$\Rightarrow \min \left\| \begin{matrix} r \times r & r \times 1 \\ D & y \end{matrix} \right\|^2 = \min \| (Dy)^T Dy \| = \min \| \underbrace{y^T D^T D}_{\substack{1 \times r & r \times r & r \times 1}} y \|^2$$

$$D^T D = \begin{bmatrix} G_0^2 & 0 & \dots & \dots \\ 0 & G_1^2 & 0 & \dots \\ \vdots & \vdots & G_2^2 & \dots \\ \vdots & \vdots & \vdots & \dots \end{bmatrix}$$

$$y^T D^T D y = [y_0 \ y_1 \ y_2] \begin{bmatrix} G_0^2 & 0 & 0 \\ 0 & G_1^2 & 0 \\ 0 & 0 & G_2^2 \end{bmatrix} \begin{bmatrix} y_0 \\ y_1 \\ y_2 \end{bmatrix}$$

$$= y_0^2 G_0^2 + y_1^2 G_1^2 + y_2^2 G_2^2, \quad G_0 > G_1 > G_2$$

let  $y = [0, 0, 1]$ , 這樣做既可以滿足  $\|y\|=1$

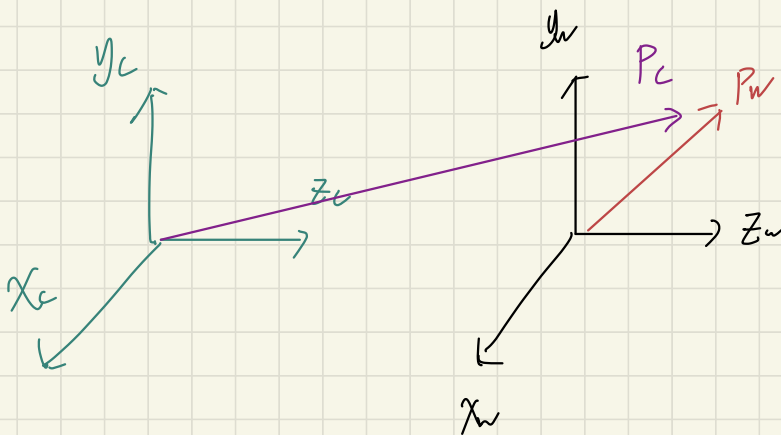
又可以讓  $\|y^T D D^T y\|$  最

分解出內、外參矩陣：

$$m = k [R | t]$$

↓                      ↓  
內參                      外參

$$= \left[ \underbrace{kR}_{3 \times 3} \mid \underbrace{kt}_{3 \times 1} \right]$$



review: 當想要將世界座標系下的一點, 轉換到相機座標系下的一點, 則:

$$P_c = [R | t] P_w$$
$$= E P_w$$

Q: 那世界座標到相機座標系的距離呢?

要先變成齊次矩陣後再求inv

$$dis = \left\| E \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \right\| = \left\| \begin{bmatrix} E^T \\ E^T \end{bmatrix} \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \right\|$$

↪ 世界座標的原點

↓ 相機座標的原點

吉文斯 (Givens) 旋轉:

利用一個旋轉矩陣將一向量剛剛好轉移到某一軸上. 即

$$R = \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix}, \quad A = \begin{bmatrix} a \\ b \end{bmatrix}$$

$$s.t. \quad RA = \begin{bmatrix} r \\ 0 \end{bmatrix}, \quad \text{令 } \cos\theta = c, \quad \sin\theta = s$$

$$\therefore \begin{bmatrix} c & -s \\ s & c \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} r \\ 0 \end{bmatrix}$$

$$\therefore R \text{ 為旋轉矩陣} \therefore \|A\| = r = \sqrt{a^2 + b^2}$$

$$\therefore \begin{cases} ca - sb = \sqrt{a^2 + b^2} & \textcircled{1} \\ sa + cb = 0 & \textcircled{2} \end{cases}$$

$$sa = -cb, \quad s = -\frac{cb}{a} \quad \textcircled{3}$$

② → ① :

$$ca + \frac{cb^2}{a} = \sqrt{a^2 + b^2}$$

$$c \left( a + \frac{b^2}{a} \right) = c \left( \frac{a^2 + b^2}{a} \right) = \sqrt{a^2 + b^2}$$

$$c = \frac{\sqrt{a^2 + b^2}}{\frac{a^2 + b^2}{a}} = \frac{a\sqrt{a^2 + b^2}}{a^2 + b^2} = \frac{a}{\sqrt{a^2 + b^2}} \quad \text{④}$$

④ → ③ :

$$s = -\frac{cb}{a} = -\frac{a}{\sqrt{a^2 + b^2}} \times \frac{b}{a} = -\frac{b}{\sqrt{a^2 + b^2}} \quad \#$$

QR 分解得出內外參:

將一矩陣分解成一上三角矩陣和一正交矩陣的內積。

$$A = RQ$$

→ 正交矩陣  
→ 上三角矩陣

ex:

$$A = \begin{bmatrix} 0 & -15 & 14 \\ 4 & 32 & 2 \\ 3 & -1 & 4 \end{bmatrix}, \text{ 利用 Givens 進行 QR 分解.}$$

想將這  
的元素都變 0

繞 x 軸:

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & c & -s \\ 0 & s & c \end{bmatrix}$$

繞 y 軸:

$$\begin{bmatrix} c & 0 & s \\ 0 & 1 & 0 \\ -s & 0 & c \end{bmatrix}$$

繞 z 軸:

$$\begin{bmatrix} c & -s & 0 \\ s & c & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

step 1: 先將  $A_{21}$  變為 0 (利用繞 z 軸)

$$R_{z1}A = \begin{bmatrix} c & -s & 0 \\ s & c & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} a & 0 & -15 & 14 \\ b & 4 & 32 & 2 \\ & 3 & -1 & 4 \end{bmatrix}$$

$$c = \frac{a}{\sqrt{a^2+b^2}} = \frac{0}{\sqrt{0^2+4^2}} = 0$$

$$s = -\frac{b}{\sqrt{a^2+b^2}} = \frac{-4}{4} = -1$$

$$R_{z1}A = \begin{bmatrix} 0 & 1 & 0 \\ -1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 0 & -15 & 14 \\ 4 & 32 & 2 \\ 3 & -1 & 4 \end{bmatrix}$$

$$= \begin{bmatrix} 4 & 32 & 2 \\ 0 & 15 & -14 \\ 3 & -1 & 4 \end{bmatrix}$$

step 2: 將  $A_{31}$  變為 0 (利用繞 y 軸)

$$R_{y1}R_{z1}A = \begin{bmatrix} c & 0 & s \\ 0 & 1 & 0 \\ -s & 0 & c \end{bmatrix} \begin{bmatrix} a & 4 & 32 & 2 \\ & 0 & 15 & -14 \\ b & 3 & -1 & 4 \end{bmatrix}$$



$$C = \frac{a}{\sqrt{a^2+b^2}} = \frac{4}{\sqrt{4^2+3^2}} = \frac{4}{5}$$

$$S = -\frac{b}{\sqrt{a^2+b^2}} = -\frac{3}{\sqrt{4^2+3^2}} = -\frac{3}{5}$$

$$\begin{aligned} R_{31}R_{21}A &= \begin{bmatrix} \frac{4}{5} & 0 & -\frac{3}{5} \\ 0 & 1 & 0 \\ -\frac{3}{5} & 0 & \frac{4}{5} \end{bmatrix} \begin{bmatrix} 4 & 32 & 2 \\ 0 & 15 & -14 \\ 3 & -1 & 4 \end{bmatrix} \\ &= \begin{bmatrix} 5 & 25 & 4 \\ 0 & 15 & -14 \\ 0 & -20 & 2 \end{bmatrix} \end{aligned}$$

step 3: 將  $A_{32}$  變為 0 (繞 X 軸)

$$R_{32}R_{31}R_{21}A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & C & -S \\ 0 & S & C \end{bmatrix} \begin{bmatrix} 5 & 25 & 4 \\ 0 & a & 15 & -14 \\ 0 & b & -20 & 2 \end{bmatrix}$$

$$C = \frac{15}{\sqrt{15^2+20^2}} = \frac{3}{5}$$

$$S = \frac{20}{\sqrt{15^2+20^2}} = \frac{4}{5}$$

$$R_{32}R_{31}R_{21}A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \frac{3}{5} & -\frac{4}{5} \\ 0 & \frac{4}{5} & \frac{3}{5} \end{bmatrix} \begin{bmatrix} 5 & 25 & 4 \\ 0 & 15 & -14 \\ 0 & -20 & 2 \end{bmatrix}$$

$$= \begin{bmatrix} 5 & 25 & 4 \\ 0 & 25 & -10 \\ 0 & 0 & -10 \end{bmatrix} = B$$

$$R_{32} R_{31} R_{21} A = B$$

$$A = R_{32}^{-1} R_{31}^{-1} R_{21}^{-1} B$$

$$= R_{32}^T R_{31}^T R_{21}^T B$$

$$= \underbrace{(R_{21} R_{31} R_{32})^T}_Q B = QR$$

↓  
正交矩陣

↓  
上三角矩陣

RQ分解:

由剛剛的QR分解得到的是正交矩陣、上三角矩陣，但我們要的是反過來的，所以需要利用以下方法。

def:  $P = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix}$ , 可得知  $P^T = P = P^{-1}$ ,

$$PP = I$$

$$\mathbf{M} = \begin{bmatrix} \mathbf{0} & m1 & m2 \\ \mathbf{0} & \mathbf{0} & m3 \\ \mathbf{0} & \mathbf{0} & \mathbf{0} \end{bmatrix}$$

則:

$$\mathbf{MP} = \begin{bmatrix} \mathbf{0} & m1 & m2 \\ \mathbf{0} & \mathbf{0} & m3 \\ \mathbf{0} & \mathbf{0} & \mathbf{0} \end{bmatrix} \begin{bmatrix} \mathbf{0} & \mathbf{0} & \mathbf{1} \\ \mathbf{0} & \mathbf{1} & \mathbf{0} \\ \mathbf{1} & \mathbf{0} & \mathbf{0} \end{bmatrix} = \begin{bmatrix} m2 & m1 & \mathbf{0} \\ m3 & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & \mathbf{0} \end{bmatrix}$$

$$\mathbf{PM} = \begin{bmatrix} \mathbf{0} & \mathbf{0} & \mathbf{1} \\ \mathbf{0} & \mathbf{1} & \mathbf{0} \\ \mathbf{1} & \mathbf{0} & \mathbf{0} \end{bmatrix} \begin{bmatrix} \mathbf{0} & m1 & m2 \\ \mathbf{0} & \mathbf{0} & m3 \\ \mathbf{0} & \mathbf{0} & \mathbf{0} \end{bmatrix} = \begin{bmatrix} \mathbf{0} & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & m3 \\ \mathbf{0} & m1 & m2 \end{bmatrix}$$

Step 1:

$$A' = PA$$

Step 2:

$$A'^T = QR$$

Step 3: 將 step 1 代入 step 2

$$(PA)^T = QR$$

$$P^T A^T = QR$$

$$A^T = QRP$$

$$A = (QRP)^T$$

$$= P^T R^T Q^T$$

$$= \underbrace{P R^T}_{\text{下三角矩陣}} \underbrace{Q^T}_{\text{正交矩陣}}$$

要將一個下三角矩陣轉為一上三角矩陣，則  
只要將  $R^T$  先左右翻再上下顛倒，

$$P R^T P : \text{上三角矩陣}$$

∴

$$A = \underbrace{(P R^T P)}_{\text{上三角矩陣}} \underbrace{(P Q^T)}_{\text{為了抵掉多出來的 } P \text{ 正交矩陣}}$$