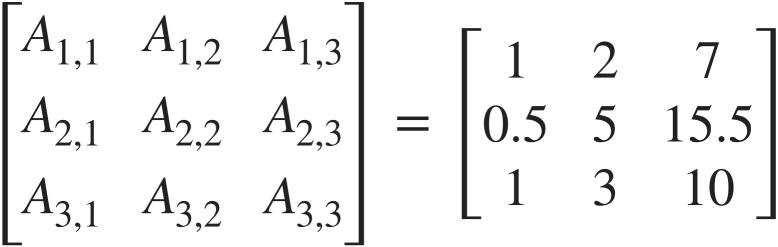
- A full rank matrix, , is one in which has N linearly independent column vectors. Two key implications arise from this:
  - Transforming the vectors space spanning , the resultant set also spans . So we say that the matrix .
  - A logical extension is that  $\boldsymbol{A}$  is a unique mapping, i.e. has a unique solution,  $\boldsymbol{x}$ , for all  $\boldsymbol{y}$  .
- Matrices with rank < N are referred to as singular.</li>

### **Matrix rank**

M-1

 $\int a_{i,k}b_{k,j}$ 

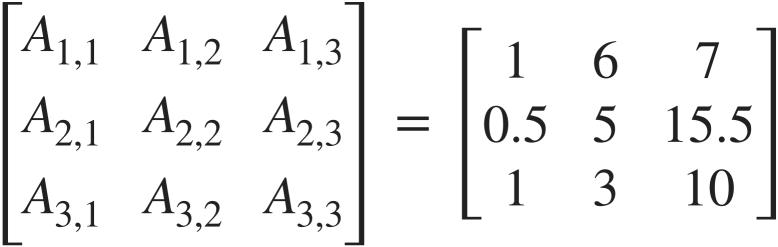
AB





### ${f A}$ spans ${\mathbb R}^N$

#### Singular matrix?



#### Singular matrix?



 $+3A._{2} =$ 

 $\Lambda$ . 2

1A. 1

#### Singular. 3rd col is linear combination of first two

#### Full rank, all columns are linearly independent

## Matrix rank

- A full rank matrix,  $\mathbf{A} \in \mathbb{R}^{N \times N}$ , is one in which has N linearly independent column vectors. Two key implications arise from this:
  - Transforming the vectors space spanning  $\mathbb{R}^N$ , the resultant set also spans  $\mathbb{R}^N$ . So we say that the matrix  $\mathbf{A}$  spans  $\mathbb{R}^N$ .
  - A logical extension is that A is a unique mapping, i.e. Ax = y has a unique solution, x, for all  $y \in \mathbb{R}^N$ .
- Matrices with rank < N are referred to as singular.</li>

Singular matrix?

$$\begin{bmatrix} A_{1,1} & A_{1,2} & A_{1,3} \\ A_{2,1} & A_{2,2} & A_{2,3} \\ A_{3,1} & A_{3,2} & A_{3,3} \end{bmatrix} = \begin{bmatrix} 1 & 2 & 7 \\ 0.5 & 5 & 15.5 \\ 1 & 3 & 10 \end{bmatrix}$$

Singular. 3rd col is linear combination of first two

$$1A_{:,1} + 3A_{:,2} = A_{:,3}$$

Singular matrix?

$$\begin{bmatrix} A_{1,1} & A_{1,2} & A_{1,3} \\ A_{2,1} & A_{2,2} & A_{2,3} \\ A_{3,1} & A_{3,2} & A_{3,3} \end{bmatrix} = \begin{bmatrix} 1 & 6 & 7 \\ 0.5 & 5 & 15.5 \\ 1 & 3 & 10 \end{bmatrix}$$

Full rank, all columns are linearly independent

# Eigendecomposition

- Definition:  $Av = \lambda v$
- Decomposition:  $\mathbf{A} = \mathbf{Q} \mathbf{\Lambda} \mathbf{Q}^T$  where  $\mathbf{Q} = \begin{bmatrix} v_1^{(1)} & \dots & v_1^{(N)} \\ \vdots & \vdots & \vdots \\ v_N^{(1)} & \dots & v_N^{(N)} \end{bmatrix}$  and  $\mathbf{\Lambda} = \begin{bmatrix} \lambda_1 & \dots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \dots & \lambda_N \end{bmatrix}$
- Applies only to square matrices

