

# Rank

## Rank of a matrix

we define the *rank* of  $A \in \mathbb{R}^{m \times n}$  as

$$\mathbf{Rank}(A) = \mathbf{dim\ range}(A)$$

(nontrivial) **facts:**

- ▶  $\mathbf{Rank}(A) = \mathbf{Rank}(A^T)$
- ▶  $\mathbf{Rank}(A)$  is maximum number of independent columns (or rows) of  $A$   
hence  $\mathbf{Rank}(A) \leq \mathbf{min}(m, n)$

## Conservation of dimension

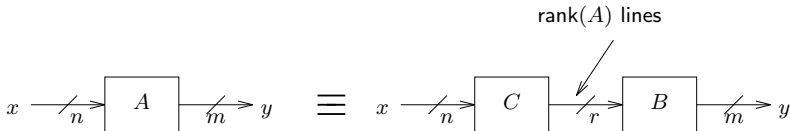
$$\mathbf{\dim \text{range}(A) + \dim \text{null}(A) = n}$$

- ▶ **Rank**( $A$ ) is dimension of set 'hit' by the mapping  $y = Ax$
- ▶ **dim null**( $A$ ) is dimension of set of  $x$  'crushed' to zero by  $y = Ax$
- ▶ 'conservation of dimension': each dimension of input is either crushed to zero or ends up in output
- ▶ roughly speaking:
  - ▶  $n$  is number of degrees of freedom in input  $x$
  - ▶ **dim null**( $A$ ) is number of degrees of freedom lost in the mapping from  $x$  to  $y = Ax$
  - ▶ **Rank**( $A$ ) is number of degrees of freedom in output  $y$

## 'Coding' interpretation of rank

$$\mathbf{Rank}(BC) \leq \min\{\mathbf{Rank}(B), \mathbf{Rank}(C)\}$$

- ▶ hence if  $A = BC$  with  $B \in \mathbb{R}^{m \times r}$ ,  $C \in \mathbb{R}^{r \times n}$ , then  $\mathbf{Rank}(A) \leq r$
- ▶ conversely: if  $\mathbf{rank}(A) = r$  then  $A \in \mathbb{R}^{m \times n}$  can be factored as  $A = BC$  with  $B \in \mathbb{R}^{m \times r}$ ,  $C \in \mathbb{R}^{r \times n}$ :
- ▶  $\mathbf{rank}(A) = r$  is minimum size vector needed to faithfully reconstruct  $y$  from  $x$



## Application: fast matrix-vector multiplication

- ▶ need to compute matrix-vector product  $y = Ax$ ,  $A \in \mathbb{R}^{m \times n}$
- ▶  $A$  has known factorization  $A = BC$ ,  $B \in \mathbb{R}^{m \times r}$
- ▶ computing  $y = Ax$  directly:  $mn$  operations
- ▶ computing  $y = Ax$  as  $y = B(Cx)$  (compute  $z = Cx$  first, then  $y = Bz$ ):  
 $rn + mr = (m + n)r$  operations
- ▶ savings can be considerable if  $r \ll \min\{m, n\}$

## Full rank matrices

for  $A \in \mathbb{R}^{m \times n}$  we always have  $\mathbf{Rank}(A) \leq \min(m, n)$

we say  $A$  is *full rank* if  $\mathbf{Rank}(A) = \min(m, n)$

- ▶ for **square** matrices, full rank means nonsingular
- ▶ for **skinny** matrices ( $m \geq n$ ), full rank means columns are independent
- ▶ for **fat** matrices ( $m \leq n$ ), full rank means rows are independent