# **Recitation 3**

#### **Rank Nullity Theorem**

#### Theorem (Rank-Nullity Theorem)

Let  $L: \mathbb{R}^m \to \mathbb{R}^n$  be a linear transformation. Then

$$\operatorname{rank}(L) + \dim(\operatorname{Ker}(L)) = m.$$

- Important theorem (check that you can reproduce the proof).
- Other things to keep in mind:
  - $rank(AB) \leq \min(rank(A), rank(B))$
  - For  $c_1, c_2, \ldots, c_m \in \mathbb{R}^n$ ,

$$\mathsf{rank}(c_1, c_2, \dots, c_m) = \mathsf{rank}(\begin{bmatrix} c_1 & c_2 & \dots & c_m \end{bmatrix}) = \mathsf{rank}\left( \begin{bmatrix} c_1 \\ c_2 \\ \dots \\ c_m \end{bmatrix} \right)$$

$$A = \begin{bmatrix} 1 & -2 & 3 \\ 0 & 1 & 0 \\ 3 & -1 & -1 \\ 4 & -2 & 0 \end{bmatrix}$$

- 1. Find a basis of the kernel of A.
- 2. Find the rank of A. Did you need to perform additional computations?
- 3. Find a basis of the image of A. Did you need to perform additional computations?

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# Symmetric Matrices

- $lacksquare A \in \mathbb{R}^{n imes n}$  symmetric if  $A_{ij} = A_{ji}$  for all  $i,j \in [1:n]$ .
- Symmetric matrices appear often and have good properties:
  - Orthogonal Projections (Lec. 4) are symmetric.
  - Spectral Theorem (Lec. 7) "symmetric matrices have an orthonormal basis of eigenvectors".
  - ▶ PCA (Lec. 7): Covariance matrix is symmetric.
  - Convexity (Lec. 9,11): Hessian Matrix (matrix of second derivative) is symmetric

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- 2. When is  $x^{\top}A^{\top}Ax = 0$ ?
- 3. Show that  $Ker(A) = Ker(A^{\top}A)$ .
- 4. Use this to show  $\operatorname{rank}(A) = \operatorname{rank}(A^{\top}A)$ .

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