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- Properties of transpose: $(A + B)^T = A^T + B^T$; $(A^T)^T = A$; $(AB)^T = B^T A^T$; $(A^{-1})^T = (A^T)^{-1}$.
- A symmetric matrix is a matrix that equals its own transpose, i.e. $A^{T} = A$.
- A symmetric matrix need *not* be invertible.
- If A is symmetric and can be factored into A = LDU without row exchanges, then U is the transpose of L. The symmetric factorization becomes $A = LDL^{T}$.
- F is a **field** if the system $(F, +, \cdot)$ satisfies the following conditions:
 - Associativity of + and ·
 - Commutativity of + and ·
 - Distributivity of \cdot over +
 - **Zero**: $\exists 0 \in F, \forall a \in F, \text{ s.t. } a + 0 = 0 + a = a$
 - $\forall a \in F, \exists -a \in F, \text{ s.t. } a + (-a) = (-a) + a = 0$
 - Unity: $\exists 1 \in F, \forall a \in F, \text{ s.t. } a \cdot 1 = 1 \cdot a = a$
 - $\bullet \ \forall a \neq 0 \in F, \exists a^{-1} \in F, \text{ s.t. } a \cdot a^{-1} = a^{-1} \cdot a = 1$
- $\mathbb{C}, \mathbb{R}, \mathbb{Q}$ are fields. \mathbb{Z}, \mathbb{N} are not.
- *V* is a **vector space** over *F* (denoted as *V/F*) if addition and scalar multiplication are defined on *V* and satisfy the algebraic rules of vector algebra:
 - Addition is associative.
 - Addition is commutative.
 - $\exists 0 \in V, \forall v \in V, \text{ s.t. } v + 0 = 0 + v = v$
 - $\forall v \in V, \exists -v \in V, \text{ s.t. } v + (-v) = (-v) + v = 0$
 - $\bullet \exists 1 \in F, \forall v \in V, \text{ s.t. } v \cdot 1 = 1 \cdot v = v$
 - $\forall v \in V, \forall \lambda, \mu \in F, \text{ s.t. } (\lambda \mu)v = \lambda(\mu v)$
 - $\forall v \in V, \forall \lambda, \mu \in F$, s.t. $(\lambda + \mu)v = \lambda v + \mu v$
 - $\forall v_1, v_2 \in V, \forall \lambda \in F, \text{ s.t. } \lambda(v_1 + v_2) = \lambda v_1 + \lambda v_2$
- Examples of vector spaces:
 - $\circ \mathbb{R}^n/\mathbb{R}$
 - $\circ \mathbb{R}^{m \times n} / \mathbb{R}$
 - Let V be the set of all symmetric matrices over \mathbb{R} . V/\mathbb{R}
 - Let V be the set of all real-valued functions defined on [0, 1]. V/\mathbb{R}
 - Let V be the set of all positive \mathbb{R} . Define x + y = xy and $cx = x^c$. V/\mathbb{R}
- S is a subspace of V over F (denoted as $S \leq V$) iff
 - S is a nonempty subset of V s.t. S itself is also a vector space over F.
 - S is closed under addition and scalar multiplication, i.e. $\forall x, y \in S \ \forall \alpha \in F \ \text{s.t.} \ \alpha x + y \in S$.
- Examples of subspaces of $\mathbb{R}^{m \times n}/\mathbb{R}$:
 - The set of all symmetric matrices.
 - The set of all upper/lower matrices.

- The set of diagonal matrices.
- The zero vector belongs to every subspace.
- If w_1 and w_2 are subspaces of V over F, then $w_1 \cap w_2 \neq \emptyset$, because $w_1 \cap w_2$ contains at least the zero vector.
- Let $V = \mathbb{R}^2/\mathbb{R}$, the followings are subspaces of V: origin, all lines through origin, V itself.
- Column space of $A \in \mathbb{R}^{m \times n}$ (denoted as $C(A) \leq \mathbb{R}^m/\mathbb{R}$) is the set of all combinations of columns of A.
- The system Ax = b is solvable iff $b \in C(A)$, i.e. $\exists x \text{ s.t. } Ax = b \text{ iff } b \in C(A)$.
- Examples of column spaces:

$$A = 0_{m \times n} \implies C(A) = 0$$

$$A = I_m \implies C(A) = \mathbb{R}^m$$

- Nullspace of $A \in \mathbb{R}^{m \times n}$ (denoted as $N(A) \leq \mathbb{R}^n / \mathbb{R}$) is $\{x \in \mathbb{R}^n \mid Ax = 0\}$
- The system Ax = 0 is called a **homogeneous equation**.
- The solution set of Ax = b is *not* a subspace of \mathbb{R}^n/\mathbb{R} .