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1 Vectors

Interpretations of Vectors

- ▷ **Vector**: an ordered list of numbers.
- ▷ Possible notations: $\vec{v} = \boldsymbol{v}$ are most common.
- ▷ **Dimensionality**: the number of the elements in a vector.
- ▷ **Geometric vector**: an object with a magnitude and direction.
- ▷ **Standard position**: when the vector begins at the origin.
- ▷ Vectors must have same dimensionality for addition and subtraction.
- ▷ **Unit vector**: a vector with a **norm** (length) of 1. Notation: $\hat{u} = \frac{\boldsymbol{u}}{|\boldsymbol{u}|}$

Vector Multiplication

- ▷ **Scalar**: scales each element in a vector, does not change direction. Generally represented with greek letters.
- ▷ **Dot product**: a single number that provides information about the relationship between two vectors. Must have **same dimensionality**.
- ▷ Notation for dot product: $\boldsymbol{a} \cdot \boldsymbol{b} = \boldsymbol{a}^T \boldsymbol{b} = \langle \boldsymbol{a} \boldsymbol{b} \rangle = \sum a_i b_i$
- ▷ *Algebraic* dot product properties:
 - **Associative**: $\boldsymbol{a}^T (\boldsymbol{b}^T \boldsymbol{c}) \neq (\boldsymbol{a}^T \boldsymbol{b})^T \boldsymbol{c}$
 - **Distributive**: $\boldsymbol{a}^T (\boldsymbol{b} + \boldsymbol{c}) = \boldsymbol{a}^T \boldsymbol{b} + \boldsymbol{a}^T \boldsymbol{c}$
 - **Commutative**: $\boldsymbol{a}^T \boldsymbol{b} = \boldsymbol{b}^T \boldsymbol{a}$
 - Vector magnitude/length: $\|\boldsymbol{v}\| = \sqrt{\boldsymbol{v}^T \boldsymbol{v}}$
- ▷ *Geometric* dot product properties:
 - Magnitudes of vectors scaled by angle between them. i.e.
 $\vec{a} = |\boldsymbol{a}| |\boldsymbol{b}| \cos(\theta_{ab})$
 - Geometric and algebraic are really the same. The above equation can be rewritten as the algebraic vector length, i.e. $\boldsymbol{a}^T \boldsymbol{b} = \cos(\theta_{ab}) |\boldsymbol{a}| |\boldsymbol{b}|$
- ▷ Dot product features based on θ :
 - If $\cos(\theta) > 0$ then $\alpha > 0$
 - If $\cos(\theta) < 0$ then $\alpha < 0$
 - If $\cos(\theta) = 0$ then $\alpha = 0$; termed **Orthogonal**
 - If $\cos(\theta) = 1$ then $\alpha = |\boldsymbol{a}| |\boldsymbol{b}|$

- ▷ *Hadamard vector multiplication*: elementwise multiplication of two vectors of equal dimensionality.
- ▷ **Outer product**: $\mathbf{vw}^T = n \times m$; a matrix resulting from the product of vectors with dimensions n and m .
- ▷ **Cross product**: defined only between two 3D vectors; produces another 3D vector that is perpendicular to both original vectors, or *normal*, to the plane containing them.
- ▷ *Complex conjugate*: the inverse sign of imaginary component of a number.
- ▷ **Hermitian transpose**: or conjugate transpose, is transpose of a vector or matrix containing imaginary numbers using the complex conjugate.
- ▷ Notation for Hermitian transpose on a matrix: \mathbf{M}^H or \mathbf{M}^*

Vector Space

- ▷ **Dimension**: in linear algebra dimension represents a *new element* of information about a vector. Geometrically, each dimension represents a *new direction*.
- ▷ **Fields**: a set of numbers on which a set of arithmetic are valid.
- ▷ Field notations:
 - \mathbb{R} : real numbers.
 - \mathbb{C} : complex numbers.
 - \mathbb{K} : real complex numbers.
 - \mathbb{Z} : integers.
 - \mathbb{N} : positive integers.
 - \mathbb{Q} : rational numbers.
 - Dimensionality in fields is written with superscripts, e.g. \mathbb{R}^N
- ▷ **Subspace**: The set of all vectors that can be created by a linear combination of some vectors or a set of vectors, i.e. $\lambda \mathbf{v}$, $\lambda \in \mathbb{R}$
 - More formal definition: a vector subspace (V) must be closed under addition and scalar multiplication, and must contain the zero vector.
 - $\forall \mathbf{v}, \mathbf{w} \in V; \forall \lambda, \alpha \in \mathbb{R}; \lambda \mathbf{v} + \alpha \mathbf{w} \in V$
 - Geometric: all possible scaled versions of vector produces a line for a single vector, and a 2d plane between two linear independent vectors.
- ▷ **Subset**: a set of points that satisfies some conditions; doesn't need to include the origin, doesn't need to be closed, and can have boundaries.

- ▷ **Span:** all possible linear combinations of vectors within a subspace.
- ▷ **Linear Independence:** a property of an entire set of vectors that is true only if no vector in the set can be written as a linear combination of the others.
 - Geometric perspective: a set of M vectors is independent if each vector points in a geometric dimension not reachable using other vectors in the set.
 - Any set of $M > N$ vectors in \mathbb{R}^N is **dependent**.
 - Any set of $M \leq N$ vectors in \mathbb{R}^N *could be independent*.
- ▷ **Basis:** a combination of span and independence. More formally: if every element of a vector space V over a field is a **linearly independent subset** of V that **spans** V .