



## Linear Algebra Cheat Sheet

V2020.10.9  
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### Matrix basic operation

- Scaling
- Matrix Addition (Subtraction)
- Matrix Multiplication

### Dot products

$$\vec{u} \cdot \vec{v} = [u_1 \ u_2] \cdot \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = u_1 v_1 + u_2 v_2$$

- Dot product measures how much two vectors move in the same direction

$$\vec{u} \cdot \vec{v} = \vec{v} \cdot \vec{u}$$

$$(c\vec{u}) \cdot \vec{v} = c(\vec{v} \cdot \vec{u})$$

$$(\vec{u} + \vec{v}) \cdot \vec{w} = \vec{u} \cdot \vec{w} + \vec{v} \cdot \vec{w}$$

$$|\vec{u} \cdot \vec{v}| \leq \|\vec{u}\| \|\vec{v}\|$$

$$\vec{u} \cdot \vec{v} = \|\vec{u}\| \|\vec{v}\| \cos \theta$$

### Linear subspace

- Basis vectors, Span
- Linear Independence
- Closed under scaling and addition

### Transformations

- Transformations are essentially functions on matrix (vector, subspace)  
 $T: \mathbb{R} \rightarrow \mathbb{R}$
- Linear Transformation as Matrix
- Reflection as Matrix
- Rotations (counterclockwise) as Matrix

$$\text{Rot}_\theta = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$$

- Rotations (3D) as Matrix

$$\text{Rot}_{\theta \text{ around } x} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos \theta & -\sin \theta \\ 0 & \sin \theta & \cos \theta \end{bmatrix}$$

$$\text{Rot}_{\theta \text{ around } y} = \begin{bmatrix} \cos \theta & 0 & \sin \theta \\ 0 & 1 & 0 \\ -\sin \theta & 0 & \cos \theta \end{bmatrix}$$

$$\text{Rot}_{\theta \text{ around } z} = \begin{bmatrix} \cos \theta & -\sin \theta & 0 \\ \sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

### Projections

- projection of a vector onto a line:

$$\text{Proj}_L(\vec{v}) = \left( \frac{\vec{v} \cdot \vec{x}}{\vec{x} \cdot \vec{x}} \right) \vec{x}$$

- projection of a vector onto a subspace (note: A may not be a square):

$$\text{Proj}_V \vec{x} = A(A^T A)^{-1} A^T \vec{x}$$

- The least square solution for  $\mathbf{Ax} = \mathbf{b}$ :

$$A^T A \vec{x}^* = A^T \vec{b}$$

- Both projections are linear

### Inverse $\mathbf{A}^{-1}$

$\mathbf{A}^{-1}$  exists,

- $\mathbf{A}^{-1}$  is unique
- $\text{Det}(\mathbf{A}) \neq 0$
- Linearly independent row/col vectors
- $\mathbf{Ax} = \mathbf{0}$  means  $\mathbf{x} = \mathbf{0}$
- If A is sparse,  $\mathbf{A}^{-1}$  could be dense

$\text{Det}(\mathbf{A}) = 0$ ,

- $\mathbf{A}^{-1}$  does not exist

### Transposes $\mathbf{A}^T$

- $(\mathbf{A} + \mathbf{B})^T = \mathbf{B}^T + \mathbf{A}^T$
- $(\mathbf{A}^T)^{-1} = (\mathbf{A}^{-1})^T$
- $(\mathbf{AB})^T = \mathbf{B}^T \mathbf{A}^T$

### Determinants

- $\text{Det}(\mathbf{A}) = |\mathbf{A}| = \begin{vmatrix} a & b \\ c & d \end{vmatrix} = ad - bc$
- If the determinant is 0, the matrix is not invertible, so there is no inverse.
- Switch any row in a matrix will change the determinant to negative
- Row operations don't change the determinant
- Geometric: parallelogram area

### Eigenvalues and Eigenvectors

- $T(\vec{v}) = \lambda \vec{v}$
- If there are  $n$  eigenvectors, they are linearly independent
- Application: Markov chain, PCA