

In The Name of God.
The Merciful, The Compassionate.

Orthogonal Vectors and Subspaces, Projecting into subspaces

notes on Gilbert Strang videos, Lecture 14-16

1 Orthogonality

- Columnspace \perp nullspace of A^T
- Orthogonal vectors: $x^T y = 0$
- Subspace \mathcal{S} is orthogonal to subspace \mathcal{T} means: every vector in \mathcal{S} is perpendicular to every vector in \mathcal{T}
- Rowspace \perp nullspace:
 $Ax = 0$, $\begin{bmatrix} \text{---} & \text{row}_1 & \text{---} \\ \text{---} & \text{row}_2 & \text{---} \\ & \vdots & \\ \text{---} & \text{row}_m & \text{---} \end{bmatrix} x = 0$, so, x is orthogonal to each row. Easy to see row combinations is also orthogonal.
- Rowspace and nullspace are orthogonal complement: nullspace contains all vectors \perp to row space.

2 Projection

- Projection of a vector b onto a line a :
 - p is the projection of b
 - e is error, $e = b - p$, $p \parallel a$
 - $\Rightarrow p = xa$
 - $a^T(b - xa) = 0 \Rightarrow xa^T a = a^T b \Rightarrow x = \frac{a^T b}{a^T a}, p = a \frac{a^T b}{a^T a}$
 - Projection Matrix = Pb , $P = \frac{aa^T}{a^T a}$
 - $C(P)$: line through a , $\text{rank}(P) = 1$
 - Two properties of projection matrices:

1. $P^T = P$

2. $P^2 = P$

- Solve $Ax = b$ when there is no solution (e.g., $m > n$):

- $A^T A$ is $n \times n$ and symmetric,
- $A^T A \hat{x} = A^T b$, \hat{x} shows estimation
- $N(A^T A) = N(A)$
- $\text{rank}(A^T A) = \text{rank}(A) \Rightarrow A^T A$ is invertible if A has independent columns
- solve $A\hat{x} = p$, p is projection of b onto columnspace
- again $e = b - p$
- $b - A\hat{x}$ is \perp to columnspace $\Rightarrow A^T(b - A\hat{x}) = 0$,
 (imagine columns of $A = \begin{bmatrix} | & | & \dots & | \\ a_1 & a_2 & \dots & a_n \\ | & | & \dots & | \end{bmatrix}$ are its bases \Rightarrow each of its columns must be \perp to e . So, $a_1^T e = 0, \dots, a_n^T e = 0 \Rightarrow A^T e = 0$)
- e is in nullspace of $A^T \Rightarrow e \perp C(A)$, YES!
- So, $A^T A \hat{x} = A^T b \Rightarrow \hat{x} = (A^T A)^{-1} A^T b$
- $p = A\hat{x} = A(A^T A)^{-1} A^T b$, $P = A(A^T A)^{-1} A^T$
- If b is in columnspace: $Pb = b \Leftarrow b$ has the form Ax , substitute by formulae of P
- If $b \perp$ columnspace: $Pb = 0 \Leftarrow b$ is in nullspace of A^T , substitute by formulae of P
- $p + e = b, p = Pb \Rightarrow e = (I - P)b$,
- $I - P$: proj to perpendicular space

- Least Squares: e.g., 2D:

- $y = C + Dt$, and we have (a_i, b_i) for i in $[1 - 3]$.

- Equations are: $C + Da_i = b_i \Rightarrow Ax = b$, $\begin{bmatrix} \vdots & \vdots \\ 1 & a_i \\ \vdots & \vdots \end{bmatrix} \begin{bmatrix} C \\ D \end{bmatrix} = \begin{bmatrix} \vdots \\ b_i \\ \vdots \end{bmatrix}$

: no solution

- Minimize: $\|Ax - b\|^2 = \|e\|^2 = \sum_i e_i^2$

- p_i are points on the line: $p + e = b$ (if p was in the place of b , the equation could be solved. So, it is in the column space of A).
- to find p , $\hat{x} = \begin{bmatrix} \hat{C} \\ \hat{D} \end{bmatrix}$:
 $A^T A \hat{x} = A^T b \rightarrow$ find $\hat{x} \rightarrow$ best line $\rightarrow p_i \rightarrow e_i$
- Note: e is perpendicular to p and columns of A
- If A has independent columns, then $A^T A$ is invertible:
 1. Suppose $A^T A x = 0$, So, x must be zero (nullspace is only the zero vector)
 2. IDEA: $x^T A^T A x = 0 \Rightarrow (Ax)^T (Ax) = 0 \Rightarrow Ax = 0$
columns of A are independent $\Rightarrow x = 0$