

- For any α , projections of b onto αa are identical.
- Given m linear equations of one variable $a_i x = b_i$, for $i = 1, \dots, m$. Let $\epsilon^2 = \sum (a_i \bar{x} - b_i)^2$.
 - If there exists a solution x s.t. $(a_1, \dots, a_m)x = (b_1, \dots, b_m)$, then $\epsilon = 0$.
 - Otherwise, there is an approximation solution \bar{x} s.t. ϵ^2 is minimized.
 - Let $\frac{d\epsilon^2}{d\bar{x}} = \sum 2a_i(a_i \bar{x} - b_i) = 0$. Then, $\bar{x} = \frac{a^\top b}{a^\top a}$.
 - \bar{x} is the coefficient s.t. $\bar{x}a$ is the projection of b onto a .
- Given m linear equations of n variables $A_{m \times n}x = b_{m \times 1}$, where $m > n$. Let $\epsilon^2 = \sum (a_i \bar{x} - b_i)^2$, where a_i 's are the rows of A for $i = 1, \dots, m$.
 - The error vector $\epsilon = A\bar{x} - b$ is perpendicular to every column of A , i.e., $A^\top(A\bar{x} - b) = 0$. Then, $A^\top A\bar{x} = A^\top b$.
 - The sum of square error (SSE) $\epsilon^2 = \|A\bar{x} - b\|^2 = (A\bar{x} - b)^\top (A\bar{x} - b)$. Let $\frac{d\epsilon^2}{d\bar{x}} = 2A^\top A\bar{x} - 2A^\top b = 0$. Then, $A^\top A\bar{x} = A^\top b$.
- The **least square solution** to an inconsistent system $Ax = b$ of m equations in n unknowns satisfies $A^\top Ax = A^\top b$, which is referred to as the **normal equations**.
- The properties of $A^\top A$:
 - Every entry of $A^\top A$ is the inner product of the i -th column and j -th column of A .
 - Symmetric. $(A^\top A)^\top = A^\top A$.
 - $A^\top A$ has the same nullspace as A .
 - If $Ax = 0$, then $A^\top Ax = 0$. Hence, $N(A) \subseteq N(A^\top A)$.
 - If $A^\top Ax = 0$, then $x^\top A^\top Ax = (Ax)^\top (Ax) = \|Ax\|^2 = 0$ iff $Ax = 0$. Hence, $N(A^\top A) \subseteq N(A)$.
 - Positive semidefinite.
- *Lemma*: If $A_{m \times n}$ has independent columns, then $A^\top A$ is nonsingular.
 - $\text{Rank}(A) = n$ and $N(A) = \{0\}$. Hence, $N(A^\top A) = \{0\}$.
- The least square solution to the inconsistent system $Ax = b$ is the solution of $A^\top Ax = A^\top b$.
 - If the columns of A are linearly independent, then $A^\top A$ is invertible. Hence, $x = (A^\top A)^{-1} A^\top b$.
 - Otherwise, $A^\top A$ is singular and $A^\top Ax = A^\top b$ has infinitely many solutions.
- The normal equation $A^\top Ax = A^\top b$ is always consistent.
- Let A be an $m \times n$ matrix over \mathbb{R} . Let $b \notin C(A)$. The closest point to b in $C(A)$ is $p = A(A^\top A)^{-1} A^\top b$. Let $P = A(A^\top A)^{-1} A^\top$.
 - P is the projection matrix that projects any vectors onto $C(A)$.
 - The column space of P is identical to the column space of A , i.e., $C(P) = C(A)$.
- An **orthogonal matrix** Q is a square matrix satisfying $Q^\top Q = I$, i.e., the columns of Q are orthonormal and $Q^{-1} = Q^\top$.
- Examples of orthogonal matrix: rotation matrix, permutation matrix.
- *Proposition*: Q preserves (1) length. $\forall x, \|Qx\| = \|x\|$. (2) inner product. $\forall x, y, \langle Qx, Qy \rangle = \langle x, y \rangle$. (3) angle. $\forall x, y, \angle(x, y) = \angle(Qx, Qy)$.