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- Let $\{q_1, \dots, q_n\}$ be orthonormal basis. $\forall b \in \mathbb{R}^n, \exists x = (x_1, \dots, x_n) \in \mathbb{R}^n, b = x_1 q_1 + \dots + x_n q_n$.
 - Let $Q = [q_1, \dots, q_n]$. Then, $x = Q^{-1}b = Q^T b$.
 - The coefficients $x_i = q_i^T b$. Hence, $b = (q_1^T b)q_1 + \dots + (q_n^T b)q_n$.
 - $(q_i^T b)q_i$ is the projection of b onto q_i .
- Let $Q_{m \times n}$ ($m > n$) be a matrix with orthogonal columns. $Q^T Q = I_{n \times n}$. Q^T is the left inverse of Q .
- Proposition:** Let $Q_{m \times n}$ ($m > n$) be a matrix with orthogonal columns. The least square solution to the normal equation $Qx = b$ is $\tilde{x} = Q^T b$. The projection matrix $P = QQ^T$.
- Norm (length) of a vector $x \in \mathbb{R}^n$: $\|x\|_1 = \sum |x_i|$. $\|x\|_2 = \sqrt{\sum x_i^2}$. $\|x\|_\infty = \max |x_i|$.
- Theorem:** Let V be an inner product space and $S = \{x_1, \dots, x_n\}$ be an orthogonal set of nonzero vectors. If $y = \sum a_i x_i$, then $a_i = \frac{\langle x_i, y \rangle}{\langle x_i, x_i \rangle}$.
- Let $\{y_1, y_2\}$ be a linearly independent set. What is the orthogonal set $\{x_1, x_2\}$ that spans the same space as $\{y_1, y_2\}$? $x_1 = y_1$. $x_2 = y_2 - \frac{\langle y_1, y_2 \rangle}{\langle y_1, y_1 \rangle} y_1 = (I - \frac{y_1 y_1^T}{y_1^T y_1}) y_2$.
- Gram-Schmidt orthogonalization process:** Let V be an inner product space and $S = \{y_1, \dots, y_n\}$ be a linearly independent set of nonzero vectors. The orthogonal set $S' = \{x_1, \dots, x_n\}$ that spans the same space as S : $x_1 = y_1$. $x_i = y_i - \sum_{k=1}^{i-1} \frac{\langle x_k, y_i \rangle}{\langle x_k, x_k \rangle} x_k$.
- Theorem:** Every $A_{m \times n}$ w/ linearly independent columns can be factored into $A = Q_{m \times n} R_{n \times n}$. The columns of Q are orthonormal and R is an invertible upper triangular matrix. When $m = n$ and all matrices are square, Q is orthogonal.
 - Let a_1, \dots, a_n be the columns of A .
 - By Gram-Schmidt orthogonalization process, we can construct orthonormal vectors q_1, \dots, q_n s.t. for $i = 1, \dots, n$, $\text{span}\{q_1, \dots, q_i\} = \text{span}\{a_1, \dots, a_i\}$.
 - For $i = 1, \dots, n$, let $q'_i = a_i - \sum_{k=1}^{i-1} (q_k^T a_i) q_k$, and $q_i = \frac{q'_i}{\|q'_i\|}$.
 - $a_i = (q_1^T a_i) q_1 + \dots + (q_{i-1}^T a_i) q_{i-1} + \|q'_i\| q_i$, which is linear combination of q_1, \dots, q_i .
 - Hence, $A = QR = [q_1, \dots, q_n] \begin{bmatrix} \|q'_1\| & \dots & q_1^T a_n \\ 0 & \ddots & \vdots \\ 0 & 0 & \|q'_n\| \end{bmatrix}$
- An inconsistent system $Ax = b$, where A has linearly independent columns, can be transformed into a consistent one. $Ax = b \rightarrow A^T Ax = A^T b \rightarrow R^T R \tilde{x} = R^T Q^T b \rightarrow R \tilde{x} = Q^T b$.
- Let $x = r \cos \theta$, $y = r \sin \theta$, $z = z$. Then, $\begin{bmatrix} dx \\ dy \\ dz \end{bmatrix} = \begin{bmatrix} \cos \theta & -r \sin \theta & 0 \\ \sin \theta & r \cos \theta & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} dr \\ d\theta \\ dz \end{bmatrix}$. Hence,

$$dV = dx dy dz = \begin{vmatrix} \cos \theta & -r \sin \theta & 0 \\ \sin \theta & r \cos \theta & 0 \\ 0 & 0 & 1 \end{vmatrix} dr d\theta dz = r dr d\theta dz.$$