

Constrained Optimization

Certain engineering problems can be converted to a formulation involving the quadratic form $Q(\underline{x}) = \underline{x}^T A \underline{x}$ subject to the constraint:

$$x_1^2 + x_2^2 + \dots + x_n^2 = 1$$

$$\text{or } \|\underline{x}\| = \|\underline{x}\|^2 = 1$$

Note: The max/min of $Q(\underline{x})$ with no cross-terms is relatively trivial to determine given the above constraint.

Example: $Q(x_1, x_2, x_3) = a x_1^2 + b x_2^2 + c x_3^2$
where $a \geq b \geq c$.

The max. can be readily found for $\|\underline{x}\| = 1$ following these steps

$$\begin{aligned} Q(\underline{x}) &= a x_1^2 + b x_2^2 + c x_3^2 \\ &\leq a x_1^2 + b x_2^2 + c x_3^2 \end{aligned}$$

$$= a (\pi_1^2 + \pi_2^2 + \pi_3^2)$$

$$= a$$

The min of $Q(\underline{x})$ can similarly be determined to be c for this example.

Theorem: If A is a symmetric matrix for the quadratic form $Q(\underline{x}) = \underline{x}^T A \underline{x}$, then

1. $\max\{Q(\underline{x})\}$ is the largest eigenvalue of A and $\max\{Q(\underline{x})\}$ is achieved when \underline{x} is the corresponding unit eigenvector.

2. $\min\{Q(\underline{x})\}$ is the smallest eigenvalue of A and $\min\{Q(\underline{x})\}$ is achieved when \underline{x} is the corresponding unit eigenvector.

Proof: Diagonalize $A \ni A = P D P^T$.

so then, $\underline{x}^T A \underline{x} = \underline{y}^T D \underline{y}$ when $\underline{x} = P \underline{y}$
 also note:

$$\begin{aligned} \|\underline{x}\|^2 &= \underline{x}^T \underline{x} = (P \underline{y})^T (P \underline{y}) \\ &= \underline{y}^T P^T P \underline{y} \\ &= \underline{y}^T I \underline{y} \\ &= \|\underline{y}\|^2 \end{aligned} \quad \text{since } P^T P = I$$

Without loss of generality, suppose
 that $A \in \mathbb{R}^{3 \times 3}$ and is diagonalized \Rightarrow

$$D = \begin{pmatrix} a & 0 & 0 \\ 0 & b & 0 \\ 0 & 0 & c \end{pmatrix}$$

where $a \geq b \geq c$, and all are
 eigenvalues of the matrix A .

It then follows from the logic
 applied in the above example that

$$1. \max \{Q(\underline{x})\} = a$$

$$2. \min \{Q(\underline{x})\} = c$$

We further note that $\underline{y}^T D \underline{y} = a$ is achieved for $\underline{y} = [1, 0, 0]^T$ by the construction of D .

The \underline{x} that then corresponds to $\max \{Q(\underline{x})\} = a$ is then determined to be,

$$\underline{x} = P \underline{y} = [\underline{v}_1 \quad \underline{v}_2 \quad \underline{v}_3] \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} = \underline{v}_1$$

where \underline{v}_1 , \underline{v}_2 , and \underline{v}_3 are the eigenvectors corresponding to a , b , and c respectively.

We therefore conclude that the $\max \{Q(\underline{x})\}$ is achieved when \underline{x} is the unit eigenvector corresponding to the largest eigenvalue of A .

Similar argument can then be applied

to the smallest eigenvalue/eigenvector.

QED