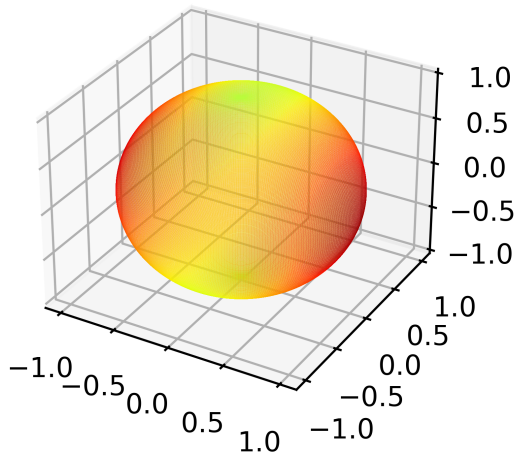


# Constrained Optimization

## An example

We have a unit sphere  $1 = x_1^2 + x_2^2 + x_3^2 = \|x\|^2$ . We wish to optimize  $Q = 9x_1^2 + 4x_2^2 + 3x_3^2$ . To find the largest and smallest value of  $Q$ . It can be graphed as follows:



We wish to maximize  $Q$ .

$$Q = 9x_1^2 + 4x_2^2 + 3x_3^2 = \vec{x}^T \begin{bmatrix} 9 & 0 & 0 \\ 0 & 4 & 0 \\ 0 & 0 & 3 \end{bmatrix} \vec{x}.$$

$$\begin{aligned} &\leq 9x_1^2 + 9x_2^2 + 9x_3^2 \\ &= 9(x_1^2 + x_2^2 + x_3^2) \\ &= 9\|\vec{x}\|^2 \\ &= 9 \end{aligned}$$

Note:  $\|\vec{x}\|^2 = 1$  because that is what we stated in the problem.

So the max value of  $Q$  is 1.

More accurately,  $\max\{Q(\vec{c}) : \|\vec{x}\| = 1\} = 9$ , and max occurs at  $\vec{x} = \begin{bmatrix} \pm 1 \\ 0 \\ 0 \end{bmatrix}$ .

If we minimize  $Q$ ,  $\min\{Q(\vec{c}) : \|\vec{x}\| = 1\} = 3$ , and min occurs at  $\vec{x} = \begin{bmatrix} 0 \\ 0 \\ \pm 1 \end{bmatrix}$ .

Notice that the minimum and maximum values of  $Q$  were the eigenvalues of  $A$ , and the corresponding eigenvectors gave their locations.

## ✎ Constrained Optimization ✓

If  $Q = x^T A x$ ,  $A$  is a real  $n \times n$  symmetric matrix, with eigenvalues

$$\lambda_1 \geq \lambda_2 \cdots \geq \lambda_n$$

and associated normalized eigenvectors

$$u_1, u_2, \dots, u_n$$

Also,  $\|x\| = 1$

Then max value of  $Q(x) = \lambda_1$  attained at  $\pm u_1$

Then min value of  $Q(x) = \lambda_n$  attained at  $\pm u_n$

### 🔗 Proof >

Assume  $\lambda_1$  is the largest eigenvalue with corresponding unit eigenvector  $\vec{u}_1$ .

$$\begin{aligned} Q &= \vec{x}^T A \vec{x} = \vec{y}^T D \vec{y}, \quad \text{using } A = P D P^T, \vec{x} = P \vec{y} \\ &= \sum \lambda_i y_i^2, \quad \text{because } D \text{ is diagonal} \\ &\leq \sum \lambda_1 y_i^2, \quad \text{because } \lambda_1 \text{ is the largest eigenvalue} \\ &= \lambda_1 \sum y_i^2 \\ &= \lambda_1 \|\vec{y}\|^2 = \lambda_1, \quad \text{because } \|\vec{y}\|^2 = 1 \end{aligned}$$

So, the maximum value of  $Q$  is at most  $\lambda_1$ . And  $Q = \lambda_1$  at  $\pm \vec{u}_1$  because

$$Q(\pm \vec{u}_1) = \vec{u}_1^T A \vec{u}_1 = \vec{u}_1^T (\lambda_1 \vec{u}_1) = \lambda_1$$