

OPTIMIZATION

Excercise 6.6

Q4. Find the equation of normal to the curve $x^2 = 4y$ which passes through the point (4,-2)

Solution: The given equation of the curve can be written as

$$g(\mathbf{x}) = \mathbf{x}^\top \mathbf{V} \mathbf{x} + 2\mathbf{u}^\top \mathbf{x} + f = 0 \quad (1)$$

where

$$\mathbf{V} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \quad (2)$$

$$\mathbf{u} = \begin{pmatrix} 0 \\ -2 \end{pmatrix} \quad (3)$$

$$f = 0 \quad (4)$$

We are given that

$$\mathbf{h} = \begin{pmatrix} 4 \\ -2 \end{pmatrix} \quad (5)$$

This can be formulated as optimization problem as follows:

$$\min_{\mathbf{x}} \quad f(\mathbf{x}) = \|\mathbf{x} - \mathbf{h}\|^2 \quad (6)$$

$$\text{s.t.} \quad g(\mathbf{x}) = \mathbf{x}^\top \mathbf{V} \mathbf{x} + 2\mathbf{u}^\top \mathbf{x} + f = 0 \quad (7)$$

First we check whether (7) is convex or not. Assume \mathbf{x}_1 and \mathbf{x}_2 that satisfy $g(\mathbf{x}) = 0$. Then,

$$g(\mathbf{x}_1) = \mathbf{x}_1^\top \mathbf{V} \mathbf{x}_1 + 2\mathbf{u}^\top \mathbf{x}_1 + f = 0 \quad (8)$$

$$g(\mathbf{x}_2) = \mathbf{x}_2^\top \mathbf{V} \mathbf{x}_2 + 2\mathbf{u}^\top \mathbf{x}_2 + f = 0 \quad (9)$$

$$(10)$$

Then, for any $0 \leq \lambda \leq 1$, we substitute

$$\mathbf{x}_\lambda = \lambda \mathbf{x}_1 + (1 - \lambda) \mathbf{x}_2 \quad (11)$$

into (7), we get

$$g(\mathbf{x}_\lambda) = (\lambda \mathbf{x}_1 + (1 - \lambda) \mathbf{x}_2)^\top \mathbf{V} (\lambda \mathbf{x}_1 + (1 - \lambda) \mathbf{x}_2) + 2\mathbf{u}^\top (\lambda \mathbf{x}_1 + (1 - \lambda) \mathbf{x}_2) + f \quad (12)$$

$$\implies (\lambda \mathbf{x}_1^\top + (1 - \lambda) \mathbf{x}_2^\top) (\lambda \mathbf{V} \mathbf{x}_1 + (1 - \lambda) \mathbf{V} \mathbf{x}_2) + 2\lambda \mathbf{u}^\top \mathbf{x}_1 + 2(1 - \lambda) \mathbf{u}^\top \mathbf{x}_2 + f \quad (13)$$

$$\begin{aligned} \implies & \lambda^2 \mathbf{x}_1^\top \mathbf{V} \mathbf{x}_1 + \lambda \mathbf{x}_2^\top \mathbf{V} \mathbf{x}_1 - \lambda^2 \mathbf{x}_2^\top \mathbf{V} \mathbf{x}_1 + \lambda \mathbf{x}_1^\top \mathbf{V} \mathbf{x}_2 + \mathbf{x}_2^\top \mathbf{V} \mathbf{x}_2 - \lambda \mathbf{x}_2^\top \mathbf{V} \mathbf{x}_2 - \lambda^2 \mathbf{x}_1^\top \mathbf{V} \mathbf{x}_2 \\ & - \lambda \mathbf{x}_2^\top \mathbf{V} \mathbf{x}_2 + \lambda^2 \mathbf{x}_2^\top \mathbf{V} \mathbf{x}_2 + 2\lambda \mathbf{u}^\top \mathbf{x}_1 + 2\mathbf{u}^\top \mathbf{x}_2 - 2\lambda \mathbf{u}^\top \mathbf{x}_2 + f \end{aligned} \quad (14)$$

Multiplying (8) by λ and (9) by $(1 - \lambda)$ and adding

$$\begin{aligned}
\lambda g(\mathbf{x}_1) + (1 - \lambda) g(\mathbf{x}_2) &= \lambda \mathbf{x}_1^\top \mathbf{V} \mathbf{x}_1 + 2\lambda \mathbf{u}^\top \mathbf{x}_1 + \lambda f + \mathbf{x}_2^\top \mathbf{V} \mathbf{x}_2 + 2\mathbf{u}^\top \mathbf{x}_2 + f \\
&\quad - \lambda \mathbf{x}_2^\top \mathbf{V} \mathbf{x}_2 + 2\lambda \mathbf{u}^\top \mathbf{x}_2 - \lambda f = 0 \\
\implies f &= -\lambda \mathbf{x}_1^\top \mathbf{V} \mathbf{x}_1 - 2\lambda \mathbf{u}^\top \mathbf{x}_1 - \mathbf{x}_2^\top \mathbf{V} \mathbf{x}_2 - 2\mathbf{u}^\top \mathbf{x}_2 \\
&\quad + \lambda \mathbf{x}_2^\top \mathbf{V} \mathbf{x}_2 - 2\lambda \mathbf{u}^\top \mathbf{x}_2 \quad (15)
\end{aligned}$$

Substituting the value of f from (15) in (12) and simplifying

$$(12) \implies (\mathbf{x}_1 - \mathbf{x}_2)^\top \mathbf{V} (\mathbf{x}_1 - \mathbf{x}_2) \quad (16)$$

Since \mathbf{V} is a semi-definite matrix, the value of (16) will be ≥ 0 contradicting the equality in (7). Hence, the optimization problem is nonconvex. However, by relaxing the constraint in (7) as

$$g(\mathbf{x}) = \mathbf{x}^\top \mathbf{V} \mathbf{x} + 2\mathbf{u}^\top \mathbf{x} + f \leq 0 \quad (17)$$

the optimization problem can be made convex. Applying convexity property to (17) and simplifying, (16) yields to

$$(\mathbf{x}_1 - \mathbf{x}_2)^\top \mathbf{V} (\mathbf{x}_1 - \mathbf{x}_2) \geq 0 \quad (18)$$

Hence the revised constraint makes it a convex optimization problem. Solving the problem using *cvxpy* we get the following point on the curve

$$\mathbf{X} = \begin{pmatrix} 1.695 \\ 0.718 \end{pmatrix} \quad (19)$$

See figure 1

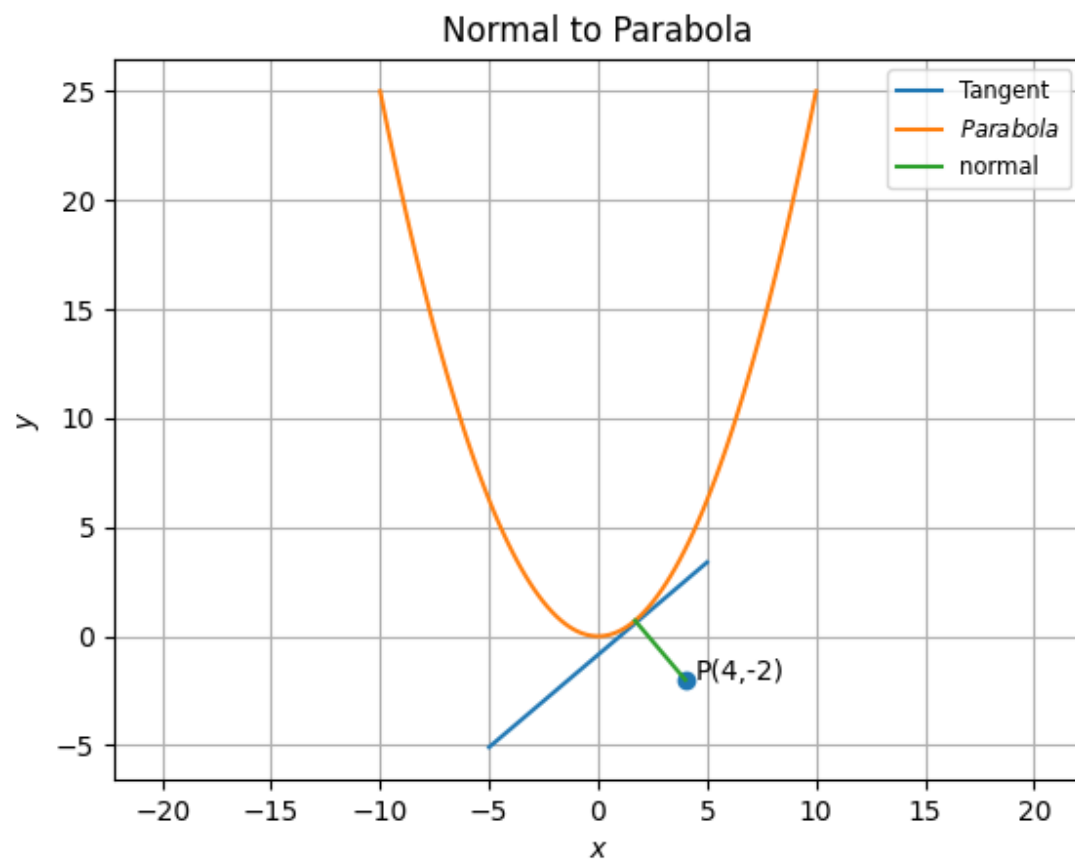


Figure 1: