



JOHNS HOPKINS

WHITING SCHOOL  
of ENGINEERING

# Algorithms for Data Science

Optimization: Quadratic Programming

# Quadratic Programming

Quadratic programming (QP) involves optimizing a quadratic objective function subject to linear constraints.

- **Standard Form:**

$$\text{Minimize } \frac{1}{2}x^T Qx + c^T x \text{ s.t. } Ax \leq b, x \geq 0$$

- **Where:**

- $Q$ : Symmetric positive semidefinite matrix, representing quadratic interactions between variables..
- $c$ : Coefficients for linear terms in the objective function.

$Ax \leq b$  Linear constraints limiting the solution space.



# QP: Mathematical Formulation

- **Objective Function:**

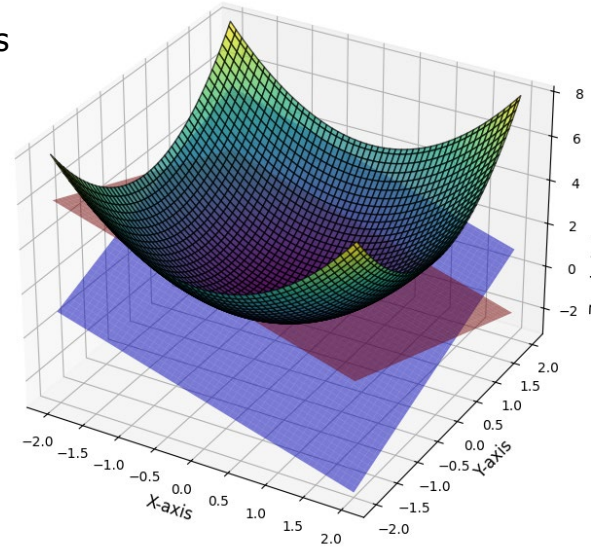
- **Quadratic Term:**  $\frac{1}{2}x^T Qx$  representing contributions of individual variables
- **Linear Term:**  $c^T x$  representing interactions between variables.

- **Constraints:**

- **Linear Inequalities:**  $Ax \leq$  limiting the solution space.
- **Non-negativity:**  $x \geq 0$  ensuring realistic solutions in practical problems.

- **Feasible Region:**

Defined by the constraints, forming a convex polyhedron for convex QP problems.



# Solving QP Problems

## Interior Point Method

- Utilizes feasible region's structure for efficient optimization.
- Operates by traversing the interior of the feasible region.

## Active-Set Method

- Iteratively solves QP problems by focusing on a subset of active constraints.

### Complexity

Convex problems can be solved in polynomial time based on their structures nature and depend on problem nize (n) and constraints (m)

- Scales efficiently for high-dimensional problems.
- Scales well for problems with sparse constraints.

# QP in Python

## Example:

$$\text{Minimize } \frac{1}{2}x_1^2 + x_2^2 - x_1x_2 - 3x_1 - 2x_2$$

s.t.

$$x_1 + x_2 \leq 2$$

$$x_1, x_2 \geq 0$$

## Mathematical Formulation:

- **Quadratic Term:**  $\frac{1}{2}x^T Q x$

$$Q = \begin{bmatrix} 1 & -0.5 \\ -0.5 & 2 \end{bmatrix}$$

- **Linear Term:**  $c^T x$

$$c = \begin{bmatrix} -3 \\ -2 \end{bmatrix}$$

- **Constraints:**

$$A = \begin{bmatrix} 1 & 1 \\ -1 & 0 \\ 0 & -1 \end{bmatrix}, b = \begin{bmatrix} 2 \\ 0 \\ 0 \end{bmatrix}$$

# QP in Python (cont.)

```
from cvxopt import matrix, solvers

# Define Q (scaled for CVXOPT) and c
Q = matrix([[1.0, -0.5], [-0.5, 2.0]]) # Scaled Q
matrix
c = matrix([-3.0, -2.0]) # Linear coefficients

# Inequality constraints Gx <= h
G = matrix([[1.0, 1.0], [-1.0, 0.0], [0.0, -1.0]], (3, 2), 'd')
h = matrix([2.0, 0.0, 0.0], (3, 1), 'd')

# Solve the QP problem
sol = solvers.qp(Q, c, G, h)

# Print the optimal solution and value
print("Optimal Solution:", sol['x'])
print("Optimal Value:", sol['primal objective'])
```

[Solving a quadratic program](#)

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