

Mixed Integer Quadratic Programming

Part I: Structural properties

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Section 1

The problem

Mixed Integer Quadratic Programming

$$\begin{aligned} \min \quad & \mathbf{x}^T Q \mathbf{x} + \mathbf{c}^T \mathbf{x} \\ \text{s. t.} \quad & A \mathbf{x} \leq \mathbf{b} \\ & \mathbf{x} \in \mathbb{Z}^p \times \mathbb{R}^{n-p} \end{aligned} \quad (\text{MIQP})$$

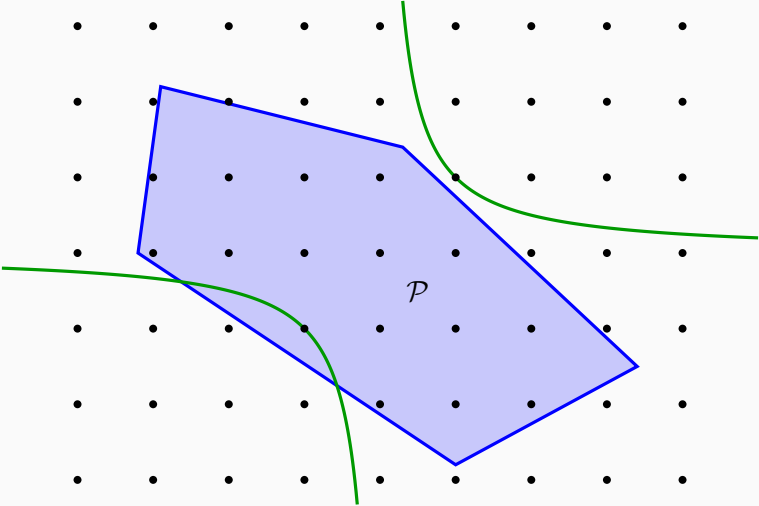
- ▶ Rational data: Q, c, A, b
- ▶ Q symmetric
- ▶ We denote by $f(\mathbf{x}) := \mathbf{x}^T Q \mathbf{x} + \mathbf{c}^T \mathbf{x}$
- ▶ We denote by $\mathcal{P} := \{\mathbf{x} \in \mathbb{R}^n \mid A \mathbf{x} \leq \mathbf{b}\}$

Mixed Integer Quadratic Programming

$$\begin{aligned} \min \quad & \mathbf{x}^T \mathbf{Q} \mathbf{x} + \mathbf{c}^T \mathbf{x} \\ \text{s. t.} \quad & \mathbf{A} \mathbf{x} \leq \mathbf{b} \\ & \mathbf{x} \in \mathbb{Z}^p \times \mathbb{R}^{n-p} \end{aligned} \quad (\text{MIQP})$$

- ▶ With $\mathbf{Q} = 0$: Mixed Integer Linear Programming (MILP)
- ▶ With $p = 0$: Quadratic Programming (QP)
- ▶ Prototypical Mixed Integer Nonlinear Programming (MINLP)

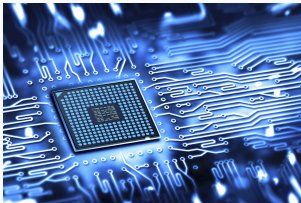
Geometry of MIQP



Applications



Power grid



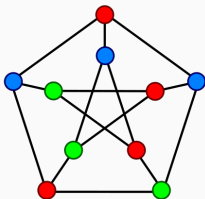
Chip design



Pooling



Game theory



Graph theory



Social choice

Application: Hydropower Scheduling

- ▶ A hydropower plant operates n turbines fed by a common water source
- ▶ **Goal:** maximize total energy production minus startup costs
- ▶ Power output of turbine i is a **quadratic function** of its flow:
 $P_i(\mathbf{x}_i) = \beta_i \mathbf{x}_i - \alpha_i \mathbf{x}_i^2$ (as flow increases, hydraulic friction losses grow **quadratically**, reducing efficiency)
- ▶ **Constraint:** each turbine is either running or shut down, with startup cost γ_i



Application: Hydropower Scheduling

Variables:

- ▶ **Continuous** $x_i \in [0, 1]$: percentage of flow through turbine i
- ▶ **Binary** $y_i \in \{0, 1\}$: turbine i is on ($y_i = 1$) or off ($y_i = 0$)

Objective function:

- ▶ $c_i = -\beta_i$: linear power output coefficient of turbine i
- ▶ $Q_{ii} = \alpha_i$: quadratic power output coefficient of turbine i
- ▶ $c_{n+i} = \gamma_i$: startup cost of turbine i (associated to y_i)

Constraints:

- ▶ **Water budget:** $\sum_{i=1}^n x_i \leq 1$
- ▶ **Linking:** $x_i \leq y_i \quad \forall i$ (no flow if turbine is off)

Application: Portfolio Optimization

- ▶ An investor allocates a budget across n assets
- ▶ **Goal:** minimize risk while maximizing expected return μ
- ▶ Risk is measured by the **variance** of the portfolio return: $\mathbf{x}^T \Sigma \mathbf{x}$
- ▶ **Constraint:** invest in at most k assets (sparse portfolio)



Application: Portfolio Optimization

Variables:

- ▶ **Continuous** $\mathbf{x}_i \in [0, 1]$: percentage of budget in asset i
- ▶ **Binary** $\mathbf{y}_i \in \{0, 1\}$: asset i is selected ($\mathbf{y}_i = 1$) or not ($\mathbf{y}_i = 0$)

Objective function:

- ▶ $Q = \Sigma$: covariance matrix of asset returns
- ▶ $c = -\mu$: negative expected returns

Constraints:

- ▶ **Budget:** $\sum_{i=1}^n \mathbf{x}_i \leq 1$
- ▶ **Linking:** $\mathbf{x}_i \leq \mathbf{y}_i \quad \forall i$
- ▶ **Cardinality:** $\sum_{i=1}^n \mathbf{y}_i \leq k$ (invest in at most k assets)

Section 2

Some fundamental properties

Some fundamental properties of MILP

$$\begin{aligned} \min \quad & c^T \mathbf{x} \\ \text{s. t.} \quad & A\mathbf{x} \leq b \\ & \mathbf{x} \in \mathbb{Z}^p \times \mathbb{R}^{n-p} \end{aligned} \quad (\text{MILP})$$

Two fundamental properties:

1. If feasible and optimal cost is not $-\infty$, there exist an **optimal solution**
2. If optimal cost is $-\infty$, there exist an **unbounded ray**

Furthermore, **optimal solutions** and **unbounded ray** can always be chosen

- ▶ **rational**, and
- ▶ of **polynomial size**

Property 1: Attainability

$$\begin{array}{ll} \inf & f(\mathbf{x}) \\ \text{s. t.} & \mathbf{x} \in \mathcal{F} \end{array}$$

Attainability:

- ▶ Assume problem feasible and optimal cost is not $-\infty$
- ▶ Does there exist an optimal solution?
- ▶ If so, we can replace \inf with \min

Some more details:

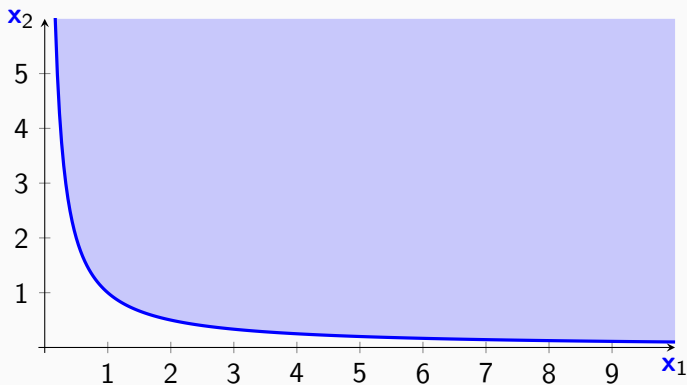
- ▶ Assume problem feasible
- ▶ Assume $\exists \gamma \in \mathbb{R}$ such that $f(\mathbf{x}) \geq \gamma, \forall \mathbf{x} \in \mathcal{F}$
- ▶ By completeness of \mathbb{R} , $\{f(\mathbf{x}) \mid \mathbf{x} \in \mathcal{F}\}$ has an infimum, say $\bar{\gamma}$
- ▶ We want to know if $\exists \bar{\mathbf{x}}$ with $f(\bar{\mathbf{x}}) = \bar{\gamma}$

Attainability does not always hold

- ▶ **False** with one quadratic inequality:

$$\begin{aligned} \inf \quad & x_1 \\ \text{s. t.} \quad & x_1 x_2 \geq 1 \\ & x_1, x_2 \geq 0 \end{aligned}$$

- ▶ Optimal cost is not $-\infty$, but no optimal solution exists

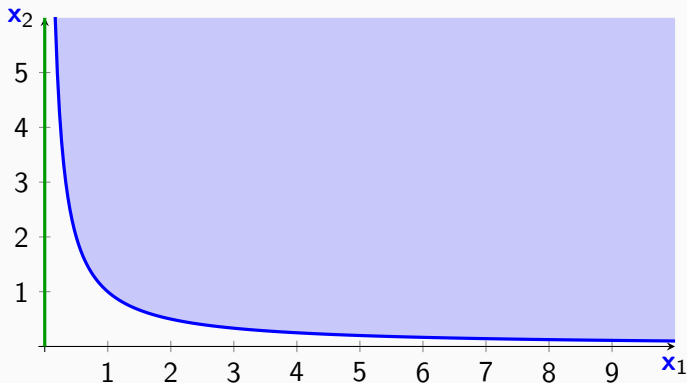


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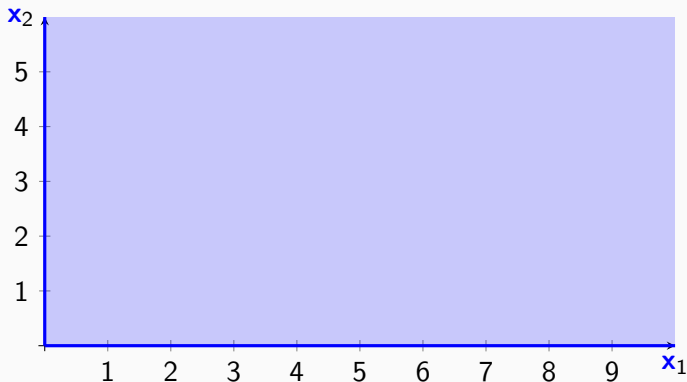


Attainability does not always hold

- ▶ **False** with quartic objective:

$$\begin{aligned} \inf \quad & x_1^2 + (x_1x_2 - 1)^2 \\ \text{s. t.} \quad & x_1, x_2 \geq 0 \end{aligned}$$

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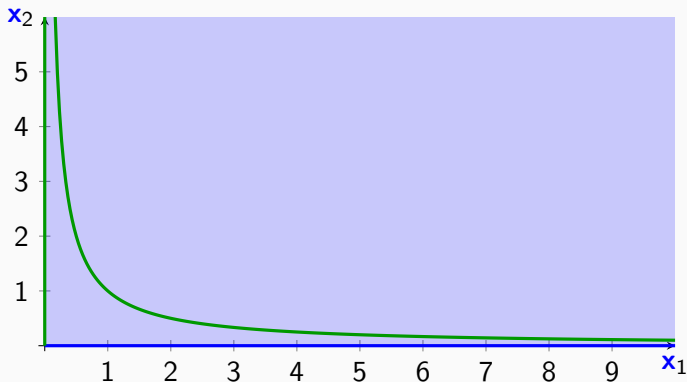


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Rationality and polynomial size

$$\begin{array}{ll} \min & f(\mathbf{x}) \\ \text{s. t.} & \mathbf{x} \in \mathcal{F} \end{array}$$

- ▶ Assume there exists an optimal solution

Rationality of optimal solutions:

- ▶ Does there exist a **rational** optimal solution?
- ▶ Meaning, an optimal solution $\bar{\mathbf{x}} \in \mathbb{Q}^n$

Optimal solutions of polynomial size:

- ▶ Does there exist an optimal solution **of polynomial size**?
- ▶ Meaning, an optimal solution $\bar{\mathbf{x}}$ of bit length polynomial in the bit length of the instance

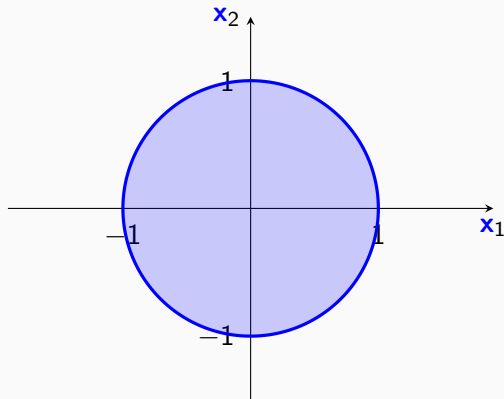
There may be no rational optimal solution

- ▶ **False** with one quadratic inequality:

$$\min \quad -x_1 - x_2$$

$$\text{s. t.} \quad x_1^2 + x_2^2 \leq 1$$

- ▶ The unique optimal solution is $(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}})$
- ▶ It is not rational



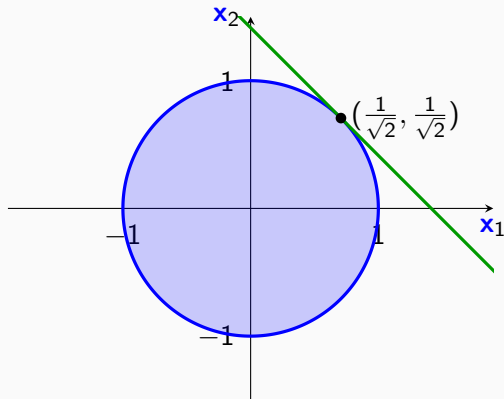
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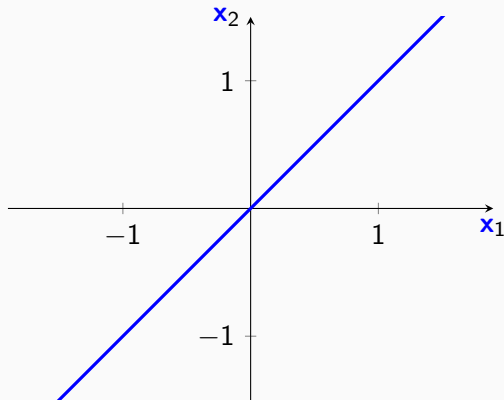
There may be no rational optimal solution

- ▶ **False** with quartic objective:

$$\min (x_1^2 + x_2^2 - 1)^2$$

$$\text{s. t. } x_1 - x_2 = 0$$

- ▶ The optimal solutions are $\pm(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}})$
- ▶ They are not rational



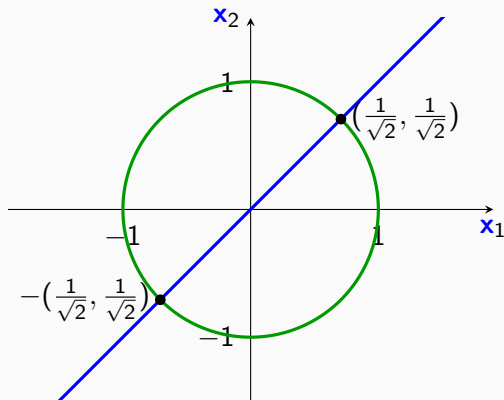
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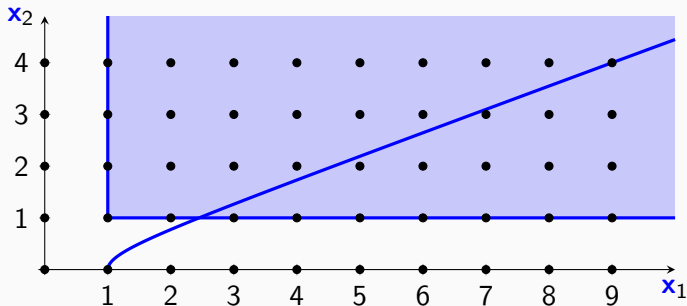


There may be no optimal solution of polynomial size

- ▶ **False** with one quadratic equality:

$$\begin{aligned} \min \quad & x_1 \\ \text{s. t.} \quad & x_1^2 - Nx_2^2 = 1 \\ & x_1, x_2 \geq 1 \\ & x_1, x_2 \in \mathbb{Z} \end{aligned}$$

- ▶ For $N = 5^{2k+1}$, $k \in \mathbb{N}$, all feasible solutions have size $\Omega(5^k)$

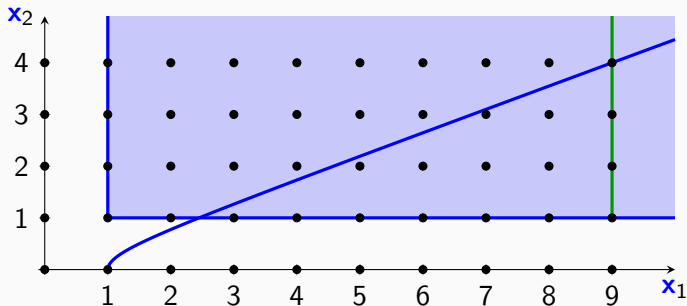


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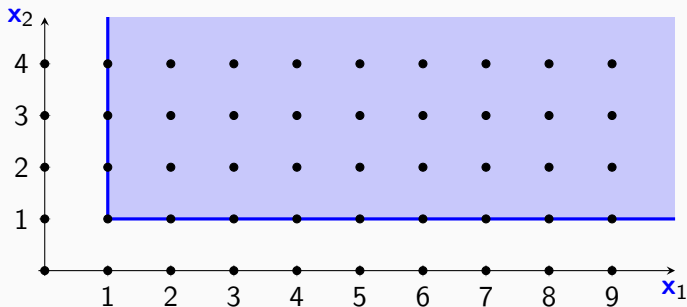
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There may be no optimal solution of polynomial size

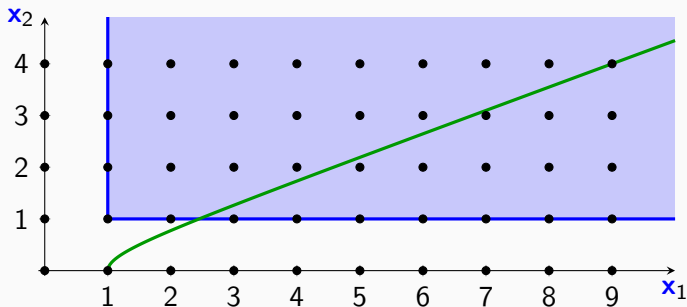
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- ▶ For $N = 5^{2k+1}$, $k \in \mathbb{N}$, all optimal solutions have size $\Omega(5^k)$



Property 2: Unbounded rays

$$\begin{array}{ll} \inf & f(\mathbf{x}) \\ \text{s. t.} & \mathbf{x} \in \mathcal{F} \end{array}$$

Unbounded rays:

- ▶ Assume optimal cost is $-\infty$
- ▶ Does there exist an **unbounded ray**?

Unbounded ray: A half line

$$\mathcal{R} := \{ \bar{\mathbf{x}} + \lambda \bar{\mathbf{d}} \mid \lambda \geq 0 \}, \quad \text{for } \bar{\mathbf{x}}, \bar{\mathbf{d}} \in \mathbb{R}^n,$$

such that the optimal cost of the problem restricted to \mathcal{R} is $-\infty$:

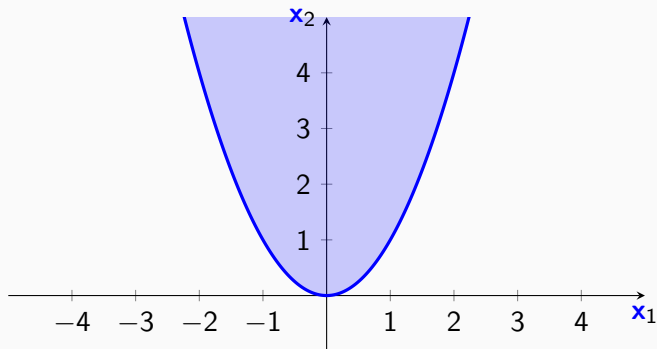
$$\begin{array}{ll} \inf & f(\mathbf{x}) \\ \text{s. t.} & \mathbf{x} \in \mathcal{F} \cap \mathcal{R} \end{array}$$

Unbounded rays may not exist

- ▶ **False** with one quadratic inequality:

$$\begin{array}{ll} \text{inf} & \mathbf{x}_1 \\ \text{s. t.} & \mathbf{x}_1^2 - \mathbf{x}_2 \leq 0 \end{array}$$

- ▶ Optimal cost is $-\infty$, but there is no unbounded ray

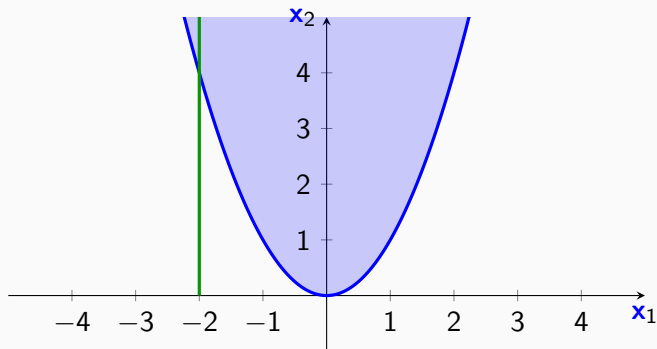


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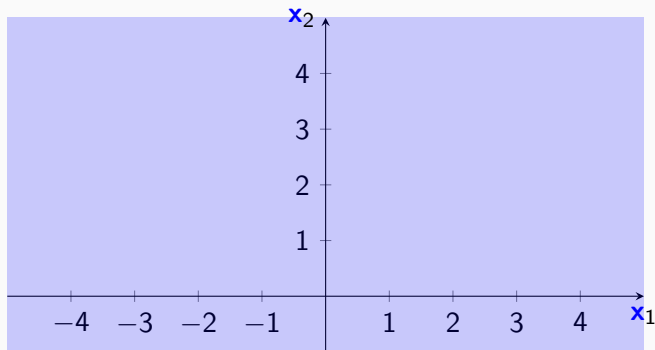


Unbounded rays may not exist

- ▶ **False** with quartic objective:

$$\begin{aligned} \inf \quad & \mathbf{x}_1 + (\mathbf{x}_1^2 - \mathbf{x}_2)^2 \\ \text{s. t.} \quad & (\mathbf{x}_1, \mathbf{x}_2) \in \mathbb{R}^2 \end{aligned}$$

- ▶ Optimal cost is $-\infty$, but there is no unbounded ray

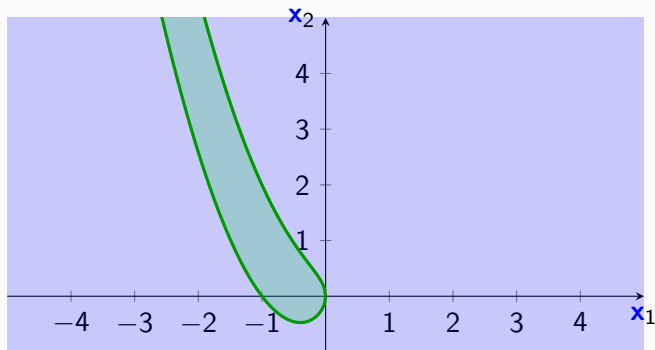


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- ▶ Optimal cost is $-\infty$, but there is no unbounded ray



Some remarks

Remark 1: All examples in **dimension two**, more general than MIQP because of:

- ▶ Either one **quadratic** constraint
- ▶ Or a **quartic** objective function

Remark 2: Why **quartic** objective functions?

- ▶ Examples easy to construct and understand
- ▶ We also know some examples with **cubic** objective functions

One more example: more quadratic inequalities

Hilbert's 10th problem [1900]:

- ▶ *It is the challenge to provide a general algorithm that, for any given Diophantine equation¹, can decide whether the equation has an integral solution*
- ▶ The solution to Hilbert's tenth problem shows that this problem is **undecidable** [Davis - Matiyasevich - Putnam - Robinson 70]
- ▶ That is, such a general algorithm cannot exist
- ▶ [Jeroslow 73] used this result to show that integer feasibility of a set defined by a fixed number of quadratic inequalities is **undecidable**
⇒ It is **not possible** to bound the size of any feasible solution

¹A polynomial equation with integer coefficients

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**WE'LL BE
BACK IN**

5

MINUTES

Section 3

A proof for MILP

Proof for MILP

We show that all properties hold for MILP:

Theorem

For any *feasible* MILP, exactly one of the following holds:

1. There is *an optimal solution* of polynomial size
2. There is *an unbounded ray* of polynomial size

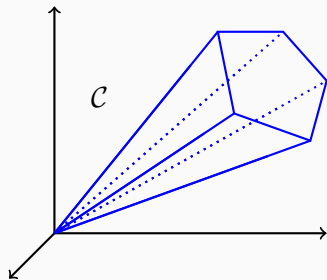
Remark: This is a proof sketch, and we focus on existence

Proof for MILP

- ▶ We can assume $\mathcal{P} \subseteq \mathbb{R}_{\geq 0}^n$
- ▶ To see that, intersect \mathcal{P} with each orthant of \mathbb{R}^n

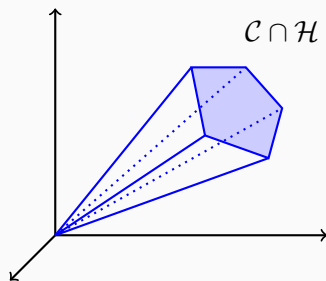
Proof for MILP

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- ▶ We consider the recession cone $\mathcal{C} := \text{rec. cone}(\mathcal{P}) \subseteq \mathbb{R}_{\geq 0}^n$



Proof for MILP

- ▶ We can assume $\mathcal{P} \subseteq \mathbb{R}_{\geq 0}^n$
- ▶ To see that, intersect \mathcal{P} with each orthant of \mathbb{R}^n
- ▶ We consider the recession cone $\mathcal{C} := \text{rec. cone}(\mathcal{P}) \subseteq \mathbb{R}_{\geq 0}^n$
- ▶ We slice it with hyperplane $\mathcal{H} := \{x \in \mathbb{R}^n \mid \sum_{i=1}^n x_i = 1\}$

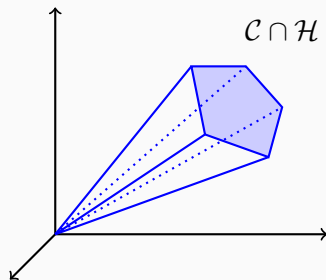


Proof for MILP

- ▶ Let $\bar{\mathbf{d}}$ be an optimal solution to the problem

$$\min \quad \mathbf{c}^T \mathbf{d}$$

$$\text{s. t.} \quad \mathbf{d} \in \mathcal{C} \cap \mathcal{H}$$

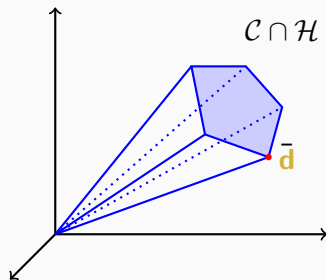


Proof for MILP

- ▶ Let $\bar{\mathbf{d}}$ be an optimal solution to the problem

$$\begin{array}{ll} \min & \mathbf{c}^T \mathbf{d} \\ \text{s. t.} & \mathbf{d} \in \mathcal{C} \cap \mathcal{H} \end{array}$$

- ▶ $\bar{\mathbf{d}}$ can be chosen as a vertex of the polytope $\mathcal{C} \cap \mathcal{H}$

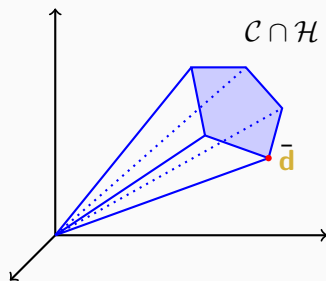


Proof for MILP

Case $c^T \bar{d} < 0$:

- ▶ Let $\bar{x} \in \mathcal{P} \cap (\mathbb{Z}^p \times \mathbb{R}^{n-p})$
- ▶ Then $\{\bar{x} + \lambda \bar{d} \mid \lambda \geq 0\}$ is an **unbounded ray**:
It contains infinitely many feasible points and

$$c^T(\bar{x} + \lambda \bar{d}) = c^T \bar{x} + \lambda \underbrace{(c^T \bar{d})}_{< 0} \rightarrow -\infty$$

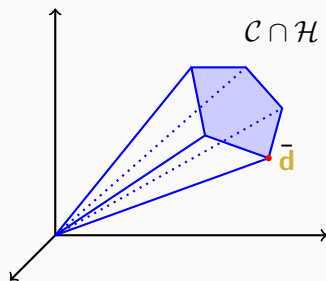


Proof for MILP

Case $c^T \bar{\mathbf{d}} \geq 0$:

► Then $c^T \mathbf{d} \geq 0$ for every $\mathbf{d} \in \mathcal{C}$, thus

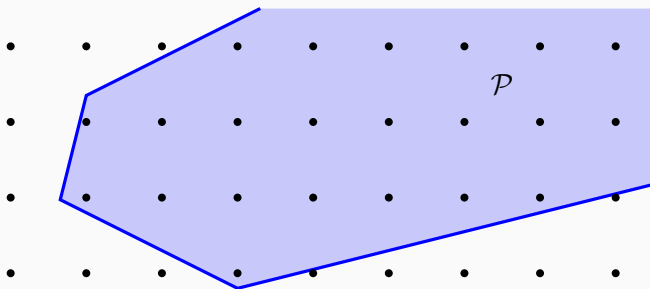
$$\begin{aligned} c^T(\mathbf{x} + \lambda \mathbf{d}) &= c^T \mathbf{x} + \lambda \underbrace{(c^T \mathbf{d})}_{\geq 0} \\ &\geq c^T \mathbf{x} \quad \forall \mathbf{x}, \forall \mathbf{d} \in \mathcal{C} \end{aligned}$$



Proof for MILP

Case $c^T \bar{\mathbf{d}} \geq 0$:

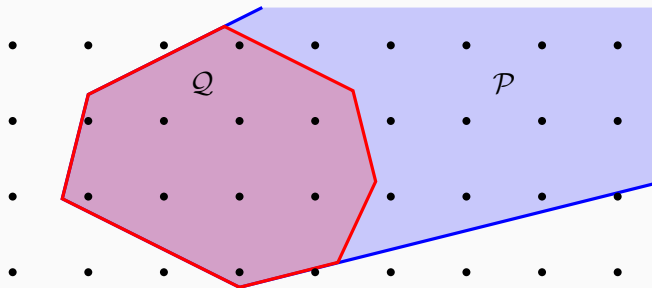
- ▶ **Fundamental Theorem of MILP:** There exists a polytope $Q \subset \mathcal{P}$ such that each feasible point is of the form $\mathbf{x} + \lambda \mathbf{d}$ with $\mathbf{x} \in Q \cap (\mathbb{Z}^p \times \mathbb{R}^{n-p})$, $\mathbf{d} \in \mathcal{C}$, $\lambda \geq 0$
- ▶ Infima over $\mathcal{P} \cap (\mathbb{Z}^p \times \mathbb{R}^{n-p})$ and $Q \cap (\mathbb{Z}^p \times \mathbb{R}^{n-p})$ coincide
- ▶ $Q \cap (\mathbb{Z}^p \times \mathbb{R}^{n-p})$ is compact, so there is an optimal solution □



Proof for MILP

Case $c^T \bar{\mathbf{d}} \geq 0$:

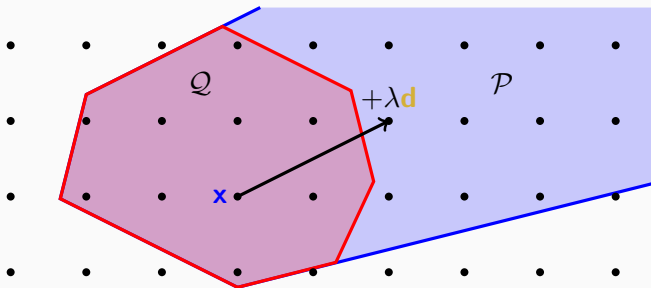
- ▶ **Fundamental Theorem of MILP**: There exists a **polytope** $Q \subset \mathcal{P}$ such that each feasible point is of the form $\mathbf{x} + \lambda \mathbf{d}$ with $\mathbf{x} \in Q \cap (\mathbb{Z}^p \times \mathbb{R}^{n-p})$, $\mathbf{d} \in \mathcal{C}$, $\lambda \geq 0$
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Proof for MILP

Case $c^T \bar{\mathbf{d}} \geq 0$:

- ▶ **Fundamental Theorem of MILP**: There exists a **polytope** $Q \subset \mathcal{P}$ such that each feasible point is of the form $\mathbf{x} + \lambda \mathbf{d}$ with $\mathbf{x} \in Q \cap (\mathbb{Z}^p \times \mathbb{R}^{n-p})$, $\mathbf{d} \in \mathcal{C}$, $\lambda \geq 0$
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- ▶ $Q \cap (\mathbb{Z}^p \times \mathbb{R}^{n-p})$ is compact, so there is an **optimal solution** □



Section 4

Two ingredients

First ingredient: QP on polytope

We need to generalize two ingredients from the MILP proof

First ingredient

- ▶ There is an optimal solution of LP on a polytope on a vertex of the polytope
- ▶ This is not true with a quadratic objective
- ▶ No problem:

Proposition

If \mathcal{P} is bounded, QP has a rational optimal solution

First ingredient: QP on polytope

Simplest case: Q positive definite, $\mathcal{P} = \mathbb{R}^n$

- ▶ The function

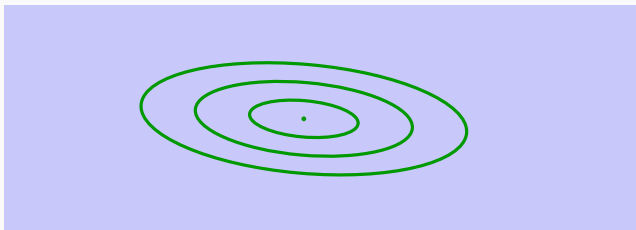
$$f(x) = \mathbf{x}^T Q \mathbf{x} + \mathbf{c}^T \mathbf{x}$$

is strictly convex, so there is a unique global minimum

- ▶ Setting the gradient to zero:

$$\nabla f(x) = 2Q\mathbf{x} + \mathbf{c} = 0 \implies \bar{\mathbf{x}} = -\frac{1}{2}Q^{-1}\mathbf{c}$$

- ▶ This is the unique global minimizer, and it is **rational**



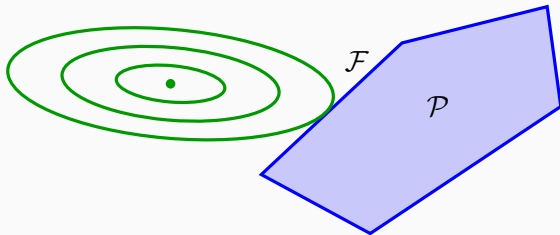
First ingredient: QP on polytope

Easy case: Q positive definite

- ▶ Since \mathcal{P} is bounded, there is an optimal solution
- ▶ Let \mathcal{F} be a face of \mathcal{P} of smallest dimension containing an optimal solution
- ▶ The optimal solution is in the interior of \mathcal{F}

$$\begin{aligned}\min \{f(\mathbf{x}) \mid \mathbf{x} \in \mathcal{P}\} &= \min \{f(\mathbf{x}) \mid \mathbf{x} \in \mathcal{F}\} \\ &= \min \{f(\mathbf{x}) \mid \mathbf{x} \in \text{aff. hull}(\mathcal{F})\}\end{aligned}$$

- ▶ We reduced the case to the previous **unconstrained case**



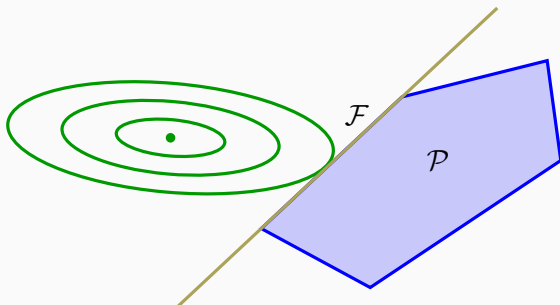
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First ingredient: QP on polytope

Remaining case: Q not positive definite

- ▶ Let $\bar{\mathbf{x}} \in \text{int}(\mathcal{P})$
- ▶ There exists $\bar{\mathbf{d}} \in \mathbb{R}^n$ such that

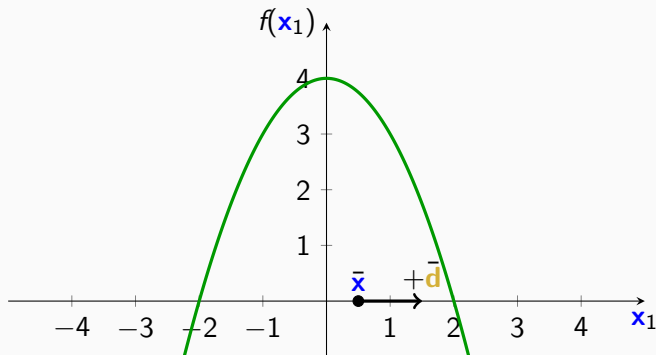
$$f(\bar{\mathbf{x}} + \lambda \bar{\mathbf{d}}) \leq f(\bar{\mathbf{x}}) \quad \forall \lambda \geq 0$$

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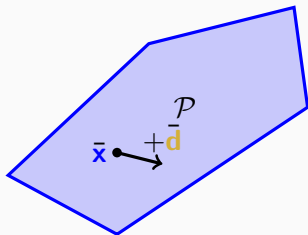
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$$f(\bar{\mathbf{x}} + \lambda \bar{\mathbf{d}}) \leq f(\bar{\mathbf{x}}) \quad \forall \lambda \geq 0$$

- ▶ Travel from $\bar{\mathbf{x}}$ in direction $\bar{\mathbf{d}}$, hit the boundary
- ▶ It suffices to consider the proper faces of \mathcal{P}



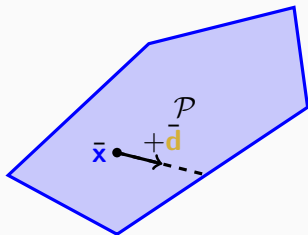
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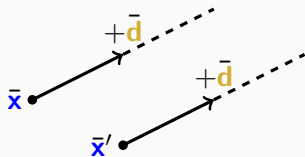
Second ingredient: ray translations

For a linear function, translating a ray maintains limit behavior

- ▶ Fix $\bar{\mathbf{d}}$ and restrict $f(\mathbf{x}) = \mathbf{c}^T \mathbf{x}$ to the ray $\{\bar{\mathbf{x}} + \lambda \bar{\mathbf{d}} \mid \lambda \geq 0\}$:

$$\begin{aligned} f(\bar{\mathbf{x}} + \lambda \bar{\mathbf{d}}) &= \mathbf{c}^T (\bar{\mathbf{x}} + \lambda \bar{\mathbf{d}}) \\ &= \mathbf{c}^T \bar{\mathbf{x}} + \lambda (\mathbf{c}^T \bar{\mathbf{d}}) \end{aligned}$$

- ▶ Linear function in λ
- ▶ Linear term: $\mathbf{c}^T \bar{\mathbf{d}}$, independent on $\bar{\mathbf{x}}$
- ▶ Constant term: $\mathbf{c}^T \bar{\mathbf{x}}$



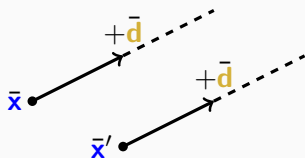
Second ingredient: ray translations

For a quadratic function, translating a ray may change limit behavior

- ▶ Fix $\bar{\mathbf{d}}$ and restrict $f(\mathbf{x}) = \mathbf{x}^T Q \mathbf{x} + \mathbf{c}^T \mathbf{x}$ to the ray $\{\bar{\mathbf{x}} + \lambda \bar{\mathbf{d}} \mid \lambda \geq 0\}$:

$$\begin{aligned} f(\bar{\mathbf{x}} + \lambda \bar{\mathbf{d}}) &= (\bar{\mathbf{x}} + \lambda \bar{\mathbf{d}})^T Q (\bar{\mathbf{x}} + \lambda \bar{\mathbf{d}}) + \mathbf{c}^T (\bar{\mathbf{x}} + \lambda \bar{\mathbf{d}}) \\ &= \lambda^2 (\bar{\mathbf{d}}^T Q \bar{\mathbf{d}}) + \lambda (2\bar{\mathbf{d}}^T Q \bar{\mathbf{x}} + \mathbf{c}^T \bar{\mathbf{d}}) + \underbrace{(\bar{\mathbf{x}}^T Q \bar{\mathbf{x}} + \mathbf{c}^T \bar{\mathbf{x}})}_{=f(\bar{\mathbf{x}})} \end{aligned}$$

- ▶ Quadratic function in λ
- ▶ Quadratic term: $\bar{\mathbf{d}}^T Q \bar{\mathbf{d}}$, independent on $\bar{\mathbf{x}}$
- ▶ Linear term: $2\bar{\mathbf{d}}^T Q \bar{\mathbf{x}} + \mathbf{c}^T \bar{\mathbf{d}}$, depends on $\bar{\mathbf{x}}$
- ▶ Constant term: $f(\bar{\mathbf{x}})$



Section 5

A proof for MIQP

Proof for MIQP

We show that all properties hold for MIQP:

Theorem

For any *feasible* MIQP, exactly one of the following holds:

1. There is *an optimal solution* of polynomial size
2. There is *an unbounded ray* of polynomial size

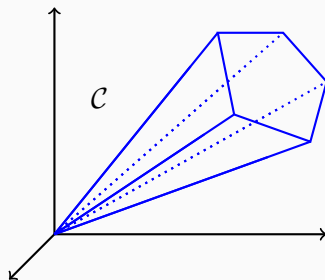
Remark: This is a proof sketch, and we focus on existence

Proof for MIQP

- ▶ We can assume $\mathcal{P} \subseteq \mathbb{R}_{\geq 0}^n$
- ▶ To see that, intersect \mathcal{P} with each orthant of \mathbb{R}^n

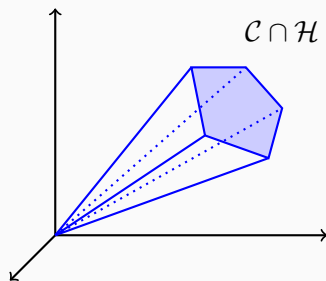
Proof for MIQP

- ▶ We can assume $\mathcal{P} \subseteq \mathbb{R}_{\geq 0}^n$
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- ▶ We consider the recession cone $\mathcal{C} := \text{rec. cone}(\mathcal{P}) \subseteq \mathbb{R}_{\geq 0}^n$



Proof for MIQP

- ▶ We can assume $\mathcal{P} \subseteq \mathbb{R}_{\geq 0}^n$
- ▶ To see that, intersect \mathcal{P} with each orthant of \mathbb{R}^n
- ▶ We consider the recession cone $\mathcal{C} := \text{rec. cone}(\mathcal{P}) \subseteq \mathbb{R}_{\geq 0}^n$
- ▶ We slice it with hyperplane $\mathcal{H} := \{x \in \mathbb{R}^n \mid \sum_{i=1}^n x_i = 1\}$

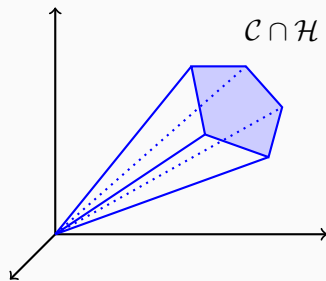


Proof for MIQP

- ▶ Let $\bar{\mathbf{d}}$ be an optimal solution to the problem

$$\min \mathbf{d}^T \mathbf{Q} \mathbf{d}$$

$$\text{s. t. } \mathbf{d} \in \mathcal{C} \cap \mathcal{H}$$

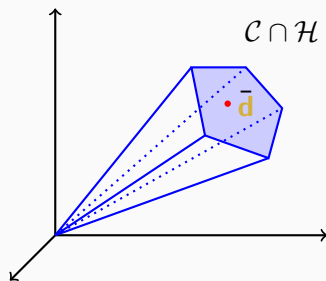


Proof for MIQP

- ▶ Let $\bar{\mathbf{d}}$ be an optimal solution to the problem

$$\begin{aligned} \min \quad & \mathbf{d}^T \mathbf{Q} \mathbf{d} \\ \text{s. t.} \quad & \mathbf{d} \in \mathcal{C} \cap \mathcal{H} \end{aligned}$$

- ▶ **First ingredient:** $\bar{\mathbf{d}}$ can be chosen rational

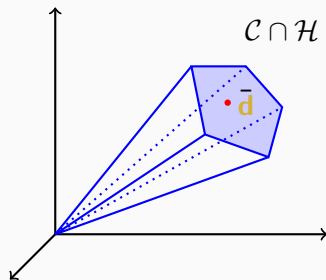


Proof for MIQP

Case $\bar{\mathbf{d}}^T Q \bar{\mathbf{d}} < 0$:

- ▶ Let $\bar{\mathbf{x}} \in \mathcal{P} \cap (\mathbb{Z}^p \times \mathbb{R}^{n-p})$
- ▶ Then $\{\bar{\mathbf{x}} + \lambda \bar{\mathbf{d}} \mid \lambda \geq 0\}$ is an **unbounded ray**:
It contains infinitely many feasible points and

$$f(\bar{\mathbf{x}} + \lambda \bar{\mathbf{d}}) = \lambda^2 \underbrace{(\bar{\mathbf{d}}^T Q \bar{\mathbf{d}})}_{< 0} + O(\lambda) \rightarrow -\infty$$

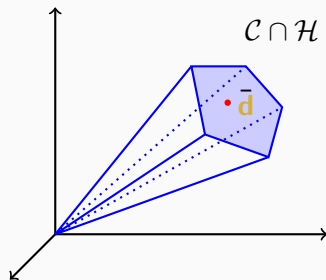


Proof for MIQP

Case $\bar{\mathbf{d}}^T Q \bar{\mathbf{d}} > 0$:

► Then $\mathbf{d}^T Q \mathbf{d} \geq \epsilon > 0$ for every $\mathbf{d} \in \mathcal{C}$, thus

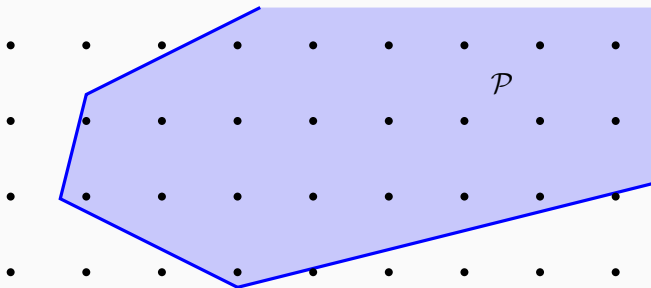
$$\begin{aligned} f(\mathbf{x} + \lambda \mathbf{d}) &= \underbrace{\lambda^2(\mathbf{d}^T Q \mathbf{d}) + \lambda(2\mathbf{d}^T Q \mathbf{x} + \mathbf{c}^T \mathbf{d})}_{\geq 0 \text{ for } \lambda \geq \bar{\lambda}} + f(\mathbf{x}) \\ &\geq f(\mathbf{x}) \quad \forall \lambda \geq \bar{\lambda}, \forall \mathbf{x}, \forall \mathbf{d} \in \mathcal{C} \end{aligned}$$



Proof for MIQP

Case $\bar{\mathbf{d}}^T Q \bar{\mathbf{d}} > 0$:

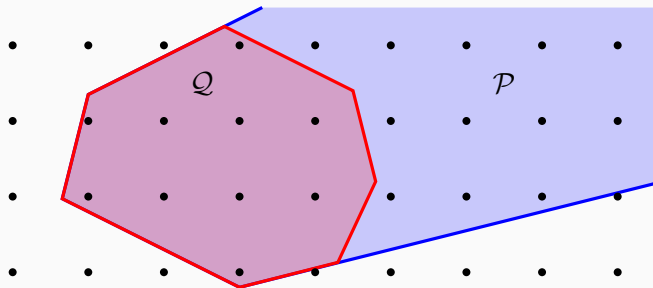
- ▶ **Fundamental Theorem of MILP:** There exists a polytope $Q \subset \mathcal{P}$ such that each feasible point is of the form $\mathbf{x} + \lambda \mathbf{d}$ with $\mathbf{x} \in Q \cap (\mathbb{Z}^p \times \mathbb{R}^{n-p})$, $\mathbf{d} \in \mathcal{C}$, $\lambda \geq 0$
- ▶ Infima over $\mathcal{P} \cap (\mathbb{Z}^p \times \mathbb{R}^{n-p})$ and $Q' \cap (\mathbb{Z}^p \times \mathbb{R}^{n-p})$ coincide, where Q' is a polytope “close” to Q
- ▶ $Q' \cap (\mathbb{Z}^p \times \mathbb{R}^{n-p})$ is compact, so there is an optimal solution



Proof for MIQP

Case $\bar{\mathbf{d}}^T Q \bar{\mathbf{d}} > 0$:

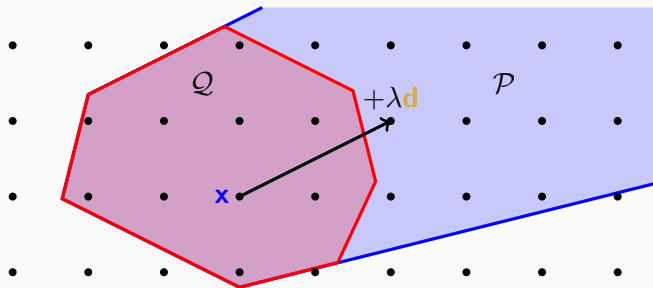
- ▶ **Fundamental Theorem of MILP:** There exists a **polytope** $Q \subset \mathcal{P}$ such that each feasible point is of the form $\mathbf{x} + \lambda \mathbf{d}$ with $\mathbf{x} \in Q \cap (\mathbb{Z}^p \times \mathbb{R}^{n-p})$, $\mathbf{d} \in \mathcal{C}$, $\lambda \geq 0$
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- ▶ $Q' \cap (\mathbb{Z}^p \times \mathbb{R}^{n-p})$ is compact, so there is an **optimal solution**



Proof for MIQP

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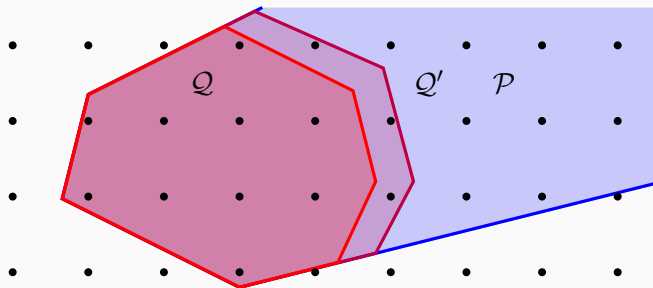
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Proof for MIQP

Case $\bar{\mathbf{d}}^T Q \bar{\mathbf{d}} > 0$:

- ▶ **Fundamental Theorem of MILP:** There exists a **polytope** $Q \subset \mathcal{P}$ such that each feasible point is of the form $\mathbf{x} + \lambda \mathbf{d}$ with $\mathbf{x} \in Q \cap (\mathbb{Z}^p \times \mathbb{R}^{n-p})$, $\mathbf{d} \in \mathcal{C}$, $\lambda \geq 0$
- ▶ Infima over $\mathcal{P} \cap (\mathbb{Z}^p \times \mathbb{R}^{n-p})$ and $Q' \cap (\mathbb{Z}^p \times \mathbb{R}^{n-p})$ coincide, where Q' is a polytope “**close**” to Q
- ▶ $Q' \cap (\mathbb{Z}^p \times \mathbb{R}^{n-p})$ is compact, so there is **an optimal solution**

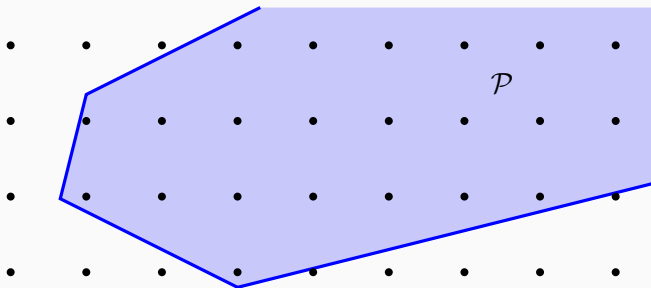


Proof for MIQP

Case $\bar{\mathbf{d}}^T \mathbf{Q} \bar{\mathbf{d}} = 0$:

- ▶ In this case the linear term $\lambda(2\bar{\mathbf{d}}^T \mathbf{Q} \mathbf{x} + c^T \bar{\mathbf{d}})$ becomes dominant, and it depends on \mathbf{x}
- ▶ Let $\bar{\mathbf{x}}$ be an optimal solution to the MILP

$$\begin{aligned} \min \quad & 2\bar{\mathbf{d}}^T \mathbf{Q} \mathbf{x} + c^T \bar{\mathbf{d}} \\ \text{s. t.} \quad & \mathbf{x} \in \mathcal{P} \cap (\mathbb{Z}^p \times \mathbb{R}^{n-p}) \end{aligned}$$

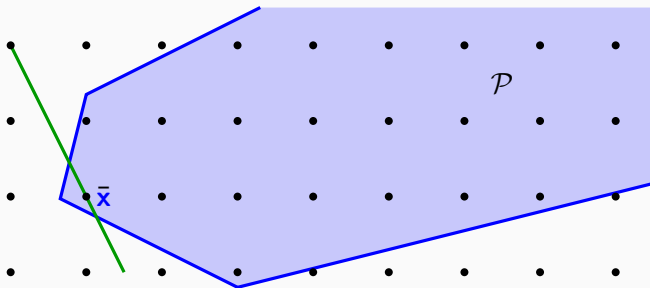


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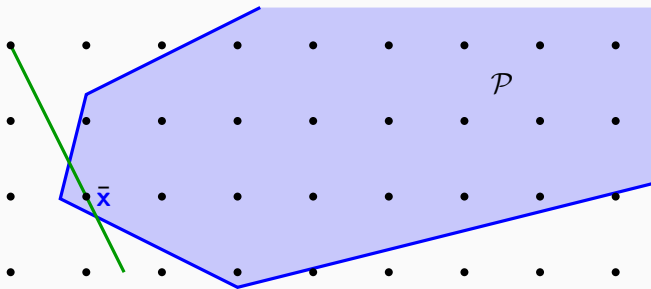
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Proof for MIQP

Sub-case $2\bar{\mathbf{d}}^T Q\bar{\mathbf{x}} + c^T\bar{\mathbf{d}} < 0$:

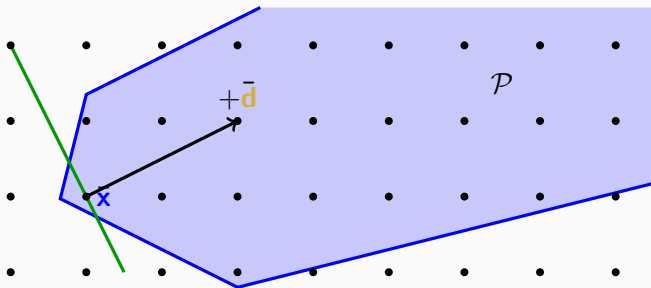
- ▶ Then $\{\bar{\mathbf{x}} + \lambda\bar{\mathbf{d}} \mid \lambda \geq 0\}$ is an **unbounded ray**



Proof for MIQP

Sub-case $2\bar{\mathbf{d}}^T \mathbf{Q}\bar{\mathbf{x}} + \mathbf{c}^T \bar{\mathbf{d}} < 0$:

- ▶ Then $\{\bar{\mathbf{x}} + \lambda \bar{\mathbf{d}} \mid \lambda \geq 0\}$ is an **unbounded ray**

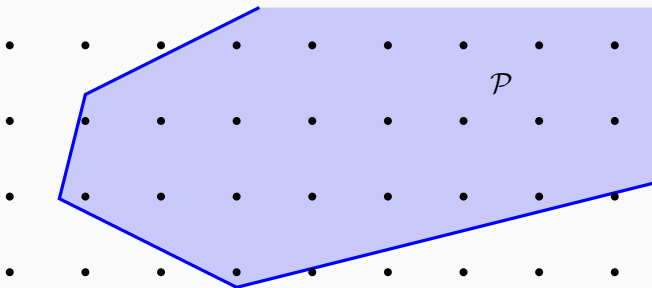


Proof for MIQP

Sub-case $2\bar{\mathbf{d}}^T \mathbf{Q}\bar{\mathbf{x}} + \mathbf{c}^T \bar{\mathbf{d}} \geq 0$:

▶ Then $2\bar{\mathbf{d}}^T \mathbf{Q}\mathbf{x} + \mathbf{c}^T \bar{\mathbf{d}} \geq 0$ for every $\mathbf{x} \in \mathcal{P} \cap (\mathbb{Z}^p \times \mathbb{R}^{n-p})$, thus

$$\begin{aligned} f(\mathbf{x} + \lambda \bar{\mathbf{d}}) &= \lambda^2 \underbrace{(\bar{\mathbf{d}}^T \mathbf{Q} \bar{\mathbf{d}})}_{=0} + \lambda \underbrace{(2\bar{\mathbf{d}}^T \mathbf{Q}\mathbf{x} + \mathbf{c}^T \bar{\mathbf{d}})}_{\geq 0} + f(\mathbf{x}) \\ &\geq f(\mathbf{x}) \quad \forall \mathbf{x} \in \mathcal{P} \cap (\mathbb{Z}^p \times \mathbb{R}^{n-p}) \end{aligned}$$

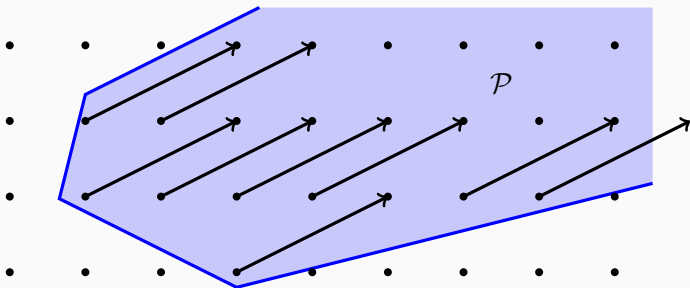


Proof for MIQP

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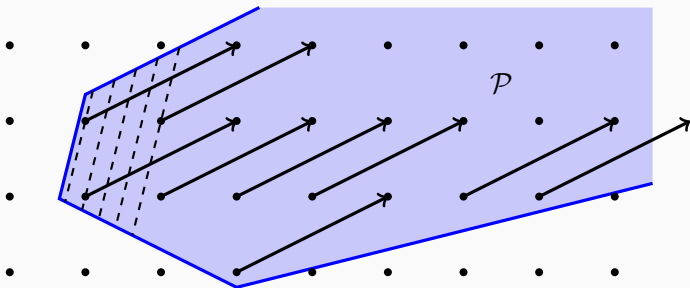
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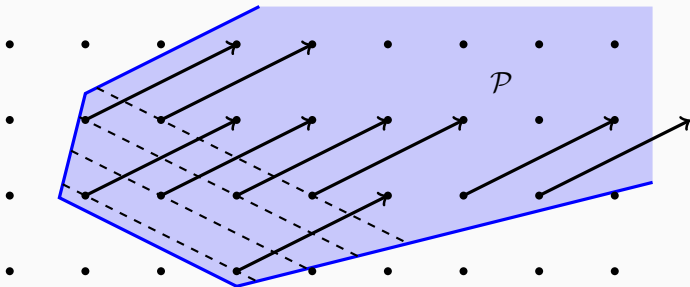
- ▶ It suffices to consider vectors “close” to the facets of \mathcal{P}
- ▶ Namely, on few translates of the facets of \mathcal{P}
- ▶ The result follows by induction on the dimension □



Proof for MIQP

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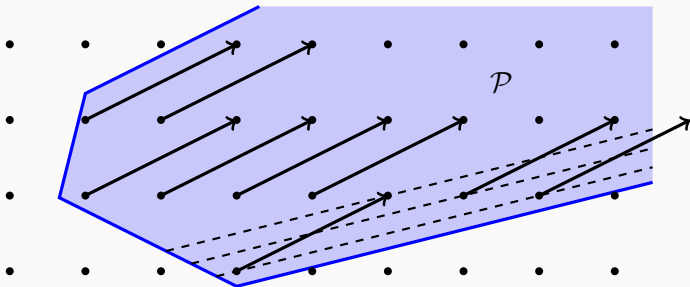
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Proof for MIQP

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- ▶ The result follows by induction on the dimension □



Enough for today!

Questions or counterexamples?