

Computer Science Track Core Course

Linear programming — recitation 1

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ISTA
Kolmogorov group

2023-05-11

Linear program

Definitions

We are given an input $A \in \mathbb{Q}^{m \times n}$, $b \in \mathbb{Q}^m$, $c \in \mathbb{Q}^n$ and we search for $\mathbf{x} \in \mathbb{Q}^n$ that optimizes...

$$\begin{array}{ll} \text{minimize} & c' \mathbf{x} \\ \text{s.t.} & A\mathbf{x} \geq b \end{array}$$

$$\begin{array}{ll} \text{maximize} & c' \mathbf{x} \\ \text{s.t.} & A\mathbf{x} \leq b \end{array}$$

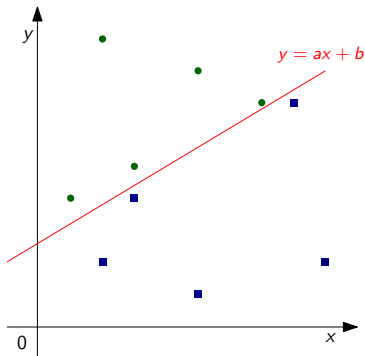
$$\begin{array}{ll} \text{minimize} & c' \mathbf{x} \\ \text{s.t.} & A\mathbf{x} = b \\ \text{where} & \mathbf{x} \geq \mathbf{0} \end{array}$$

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Example of LP

Separating points

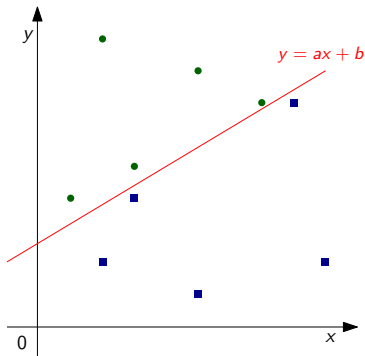
We are searching for a line $y = ax + b$ that separates “disks” (top) from “squares” (bottom).



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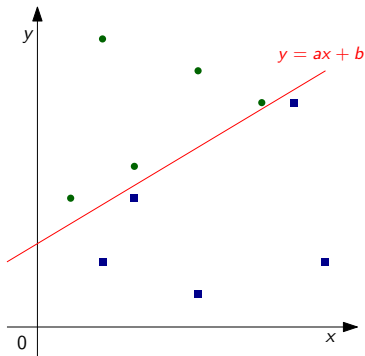


Variables:

Example of LP

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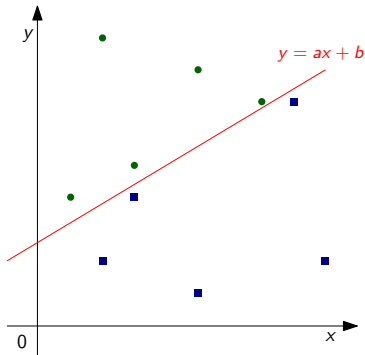


Variables: $a \leq 0$, $b \leq 0$

Example of LP

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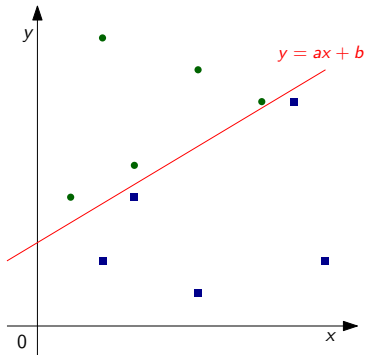
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Point (x_1, y_1) above the line:

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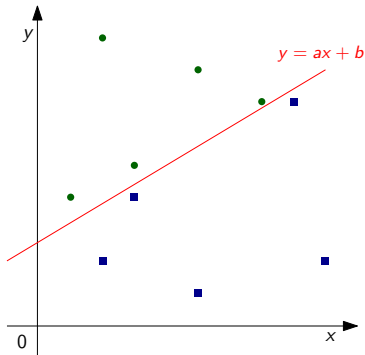
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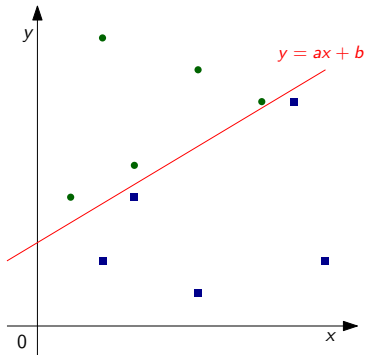
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Example of LP

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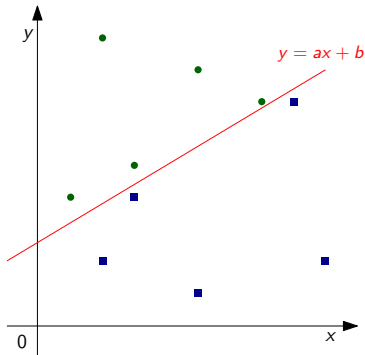
Point (x_2, y_2) below the line: $x_2 \cdot a + b \geq y_2$

“square”

Example of LP

Separating points

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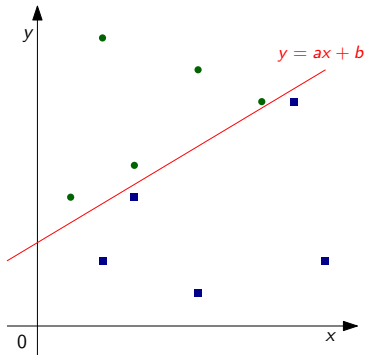
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Objective:

Example of LP

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We are searching for a line $y = ax + b$ that separates “disks” (top) from “squares” (bottom).



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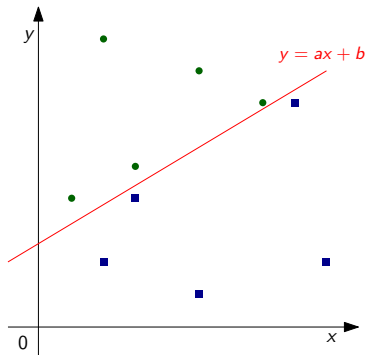
“square”

Objective: minimize 0

Example of LP

Separating points strictly

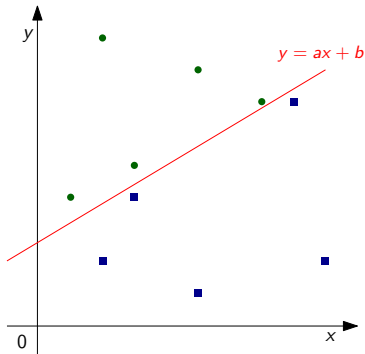
Now, we are searching for a line $y = ax + b$ that separates them strictly.



Example of LP

Separating points strictly

Now, we are searching for a line $y = ax + b$ that separates them strictly.



Variables: $a \leq 0$, $b \leq 0$, $d \geq 0$

Point (x_1, y_1) above the line: $x_1 \cdot a + b + d \leq y_1$

Point (x_2, y_2) below the line: $x_2 \cdot a + b - d \geq y_2$

Objective: maximize d

Integer linear program

Definitions

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Example of ILP

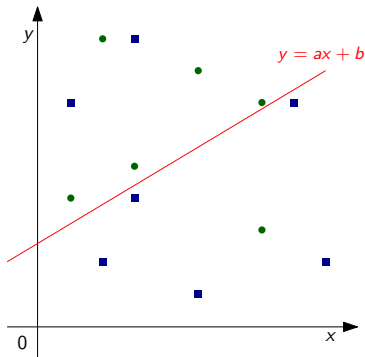
Separating points partially

We are searching for a line $y = ax + b$ that separates given points — “disks” (top) from “squares” (bottom) — with as few exceptions as possible.

Let us ignore all vertical and nearly-vertical solutions.

M is a very large integer.

Variables:



Example of ILP

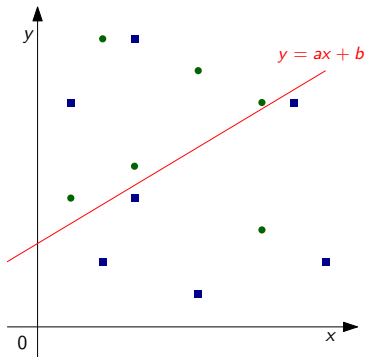
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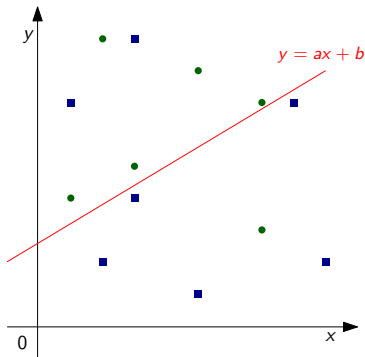
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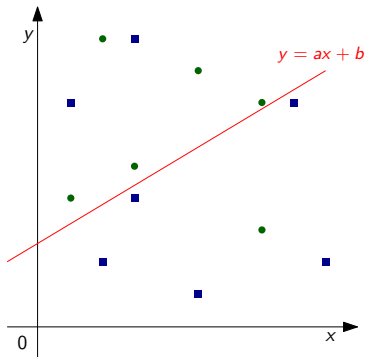
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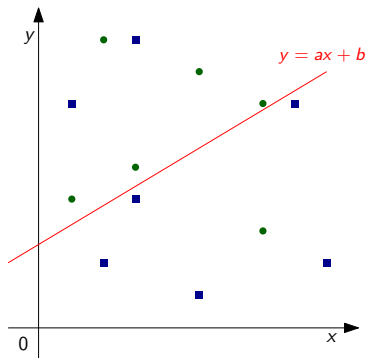
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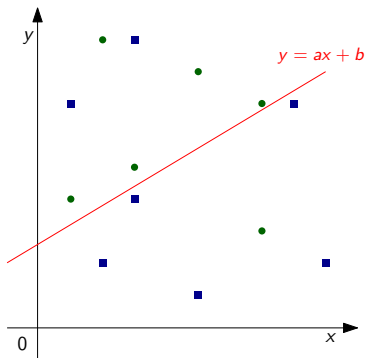
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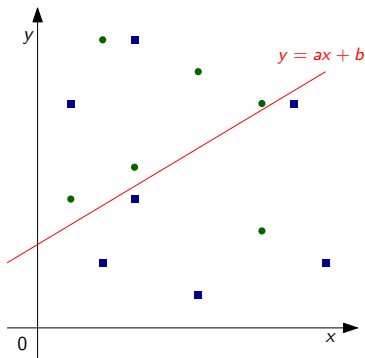
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Objective:



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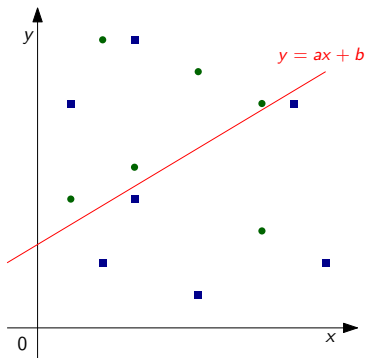
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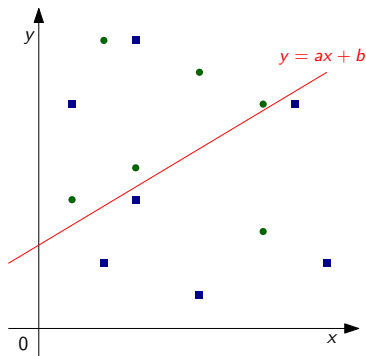
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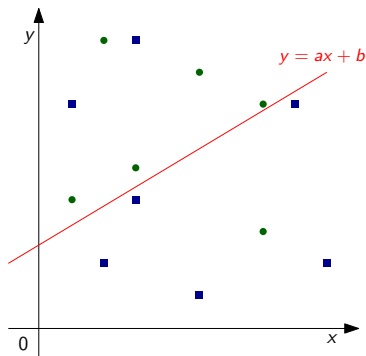
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Unfortunately, there is no general algorithm for solving ILP in polynomial time

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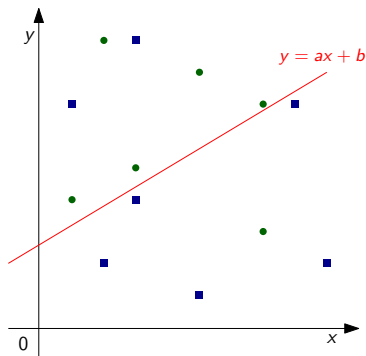
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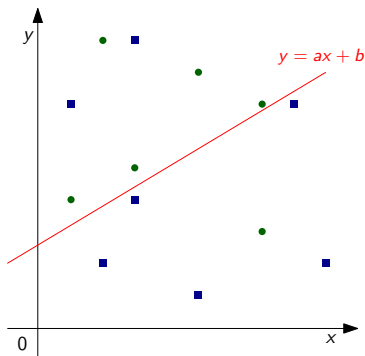
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Unfortunately, there is no general algorithm for solving ILP in polynomial time (we don't know any such algorithm — but most importantly, it cannot exist unless $P = NP$).

This particular problem, however, can be solved in quadratic time. Consider all pairs of points.

Example of LP duality

From minimization to maximization

Primal problem:

$$\begin{array}{llllll} \text{minimize} & 6x_1 & + & 6x_2 & & \\ \text{s.t.} & 2x_1 & + & x_2 & \geq & 4 \\ & x_1 & + & 2x_2 & \geq & 5 \\ \text{where} & x_1, x_2 & \geq & 0 & & \end{array}$$

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Example of LP duality

Comparison

Primal problem:

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LP duality

In general (I)

We are given $A \in \mathbb{Q}^{m \times n}$, $b \in \mathbb{Q}^m$, $c \in \mathbb{Q}^n$ on the input.
We search for either $x \in \mathbb{Q}^n$ or $y \in \mathbb{Q}^m$.

Primal problem:

$$\begin{array}{ll}\text{minimize} & c'x \\ \text{s.t.} & Ax \geq b \\ \text{where} & x \leq 0\end{array}$$

Dual problem:

$$\begin{array}{ll}\text{maximize} & b'y \\ \text{s.t.} & A^T y = c \\ \text{where} & y \geq 0\end{array}$$

LP duality

In general (II)

We are given $A \in \mathbb{Q}^{m \times n}$, $b \in \mathbb{Q}^m$, $c \in \mathbb{Q}^n$ on the input.
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LP duality

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Combinations of vectors (linear, affine, convex)

Definition

Let V be a vector space over F . Consider vectors $\mathbf{x}_1, \dots, \mathbf{x}_n \in V$.

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- We say that $\mathbf{y} \in V$ is a **linear combination** of $\mathbf{x}_1, \dots, \mathbf{x}_n$ if:

$$\exists \alpha_1, \dots, \alpha_n \in F : \sum_{i=1}^n \alpha_i \mathbf{x}_i = \mathbf{y}$$

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- Suppose now that F is a totally ordered field.

Denote the set of $\varphi \in F$ such that $\varphi \geq 0$ by the symbol F_0^+ .

We say that $\mathbf{y} \in V$ is a **convex combination** of $\mathbf{x}_1, \dots, \mathbf{x}_n$ if:

$$\exists \alpha_1, \dots, \alpha_n \in F_0^+ : \sum_{i=1}^n \alpha_i = 1 \wedge \sum_{i=1}^n \alpha_i \mathbf{x}_i = \mathbf{y}$$

Combinations of vectors (linear, affine, convex)

Quiz

Let $d \in \mathbb{Z}$ such that $d \geq 2$.

What is the **maximum** possible amount (largest set) of:

- **Linearly** independent vectors in \mathbb{Z}_2^d over \mathbb{Z}_2 ?
- **Affinely** independent vectors in \mathbb{C}^d over \mathbb{C} ?
- **Convexly** independent vectors in \mathbb{Q}^d over \mathbb{Q} ?

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 $d + 1$
- **Convexly** independent vectors in \mathbb{Q}^d over \mathbb{Q} ?

Combinations of vectors (linear, affine, convex)

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- **Linearly** independent vectors in \mathbb{Z}_2^d over \mathbb{Z}_2 ?
 d
- **Affinely** independent vectors in \mathbb{C}^d over \mathbb{C} ?
 $d + 1$
- **Convexly** independent vectors in \mathbb{Q}^d over \mathbb{Q} ?
 ∞

Combinations of vectors (linear, affine, convex)

Basic properties

$$\{\text{convex combin.}\} \subseteq \{\text{affine combin.}\} \subseteq \{\text{linear combin.}\}$$

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Let us have a matrix $A \in F^{m \times n}$ and a vector $b \in F^m$.

We search for a solution $x \in F^n$.

- $Ax = 0$: Any linear combination of solutions is a solution.
- $Ax = b$: Any affine combination of solutions is a solution.
- $Ax \leq b$: Any convex combination of solutions is a solution.

In this example, F must be a totally ordered field.