

# Iterated Linear Optimization

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# Overview

Discrete dynamical system defined by repeated linear optimization.

Compact convex set  $\Delta \subseteq \mathbb{R}^n$ ,  $T : \Delta \rightarrow \Delta$ ,

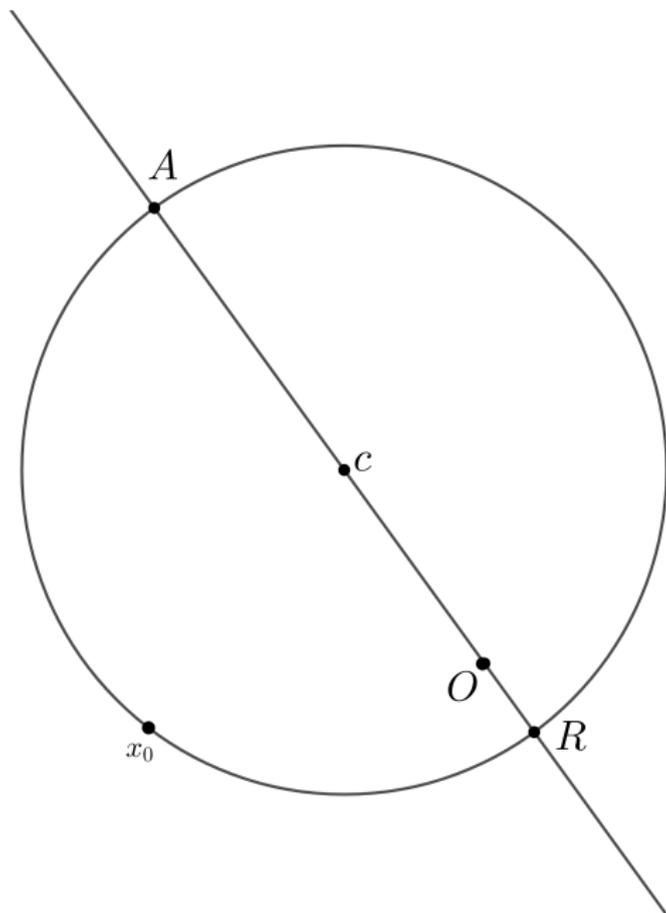
$$T(x) = \operatorname{argmax}_{y \in \Delta} (x \cdot y)$$

Fixed point iteration generates a sequence  $\{x_0, x_1, x_2, \dots\}$  where

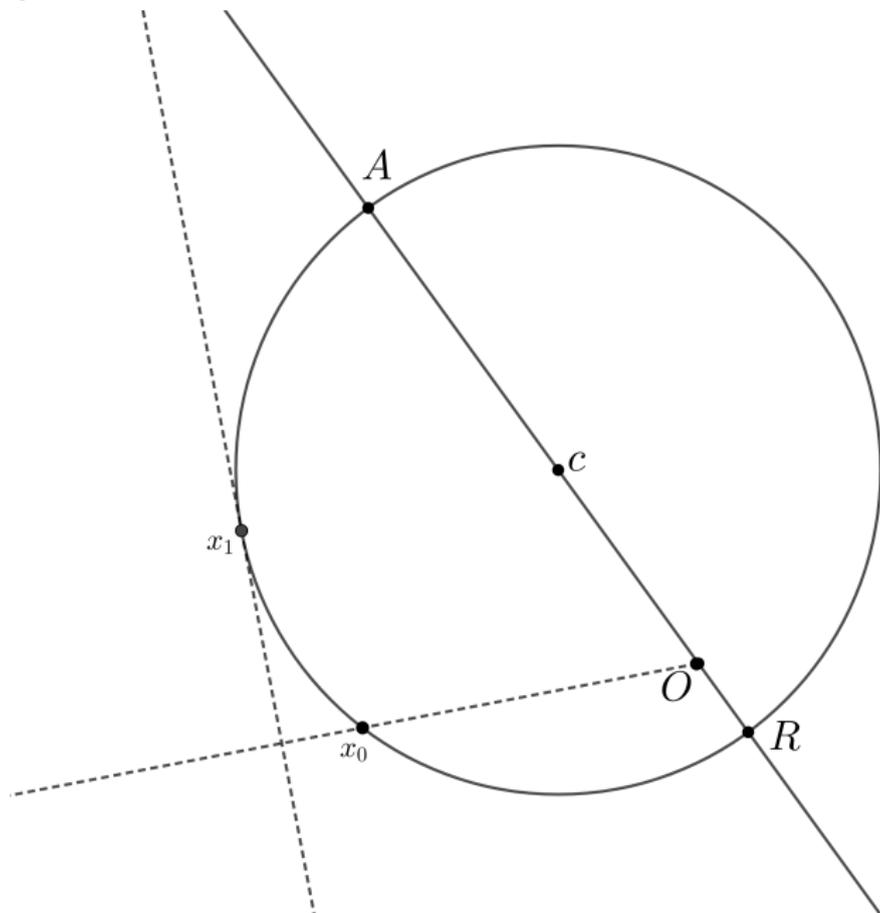
$$x_{t+1} = T(x_t)$$

- Fixed points reflect geometric properties of  $\Delta$ .
- Can be used for rounding solutions of semidefinite relaxations.
- We characterize the fixed points in *elliptopes*.

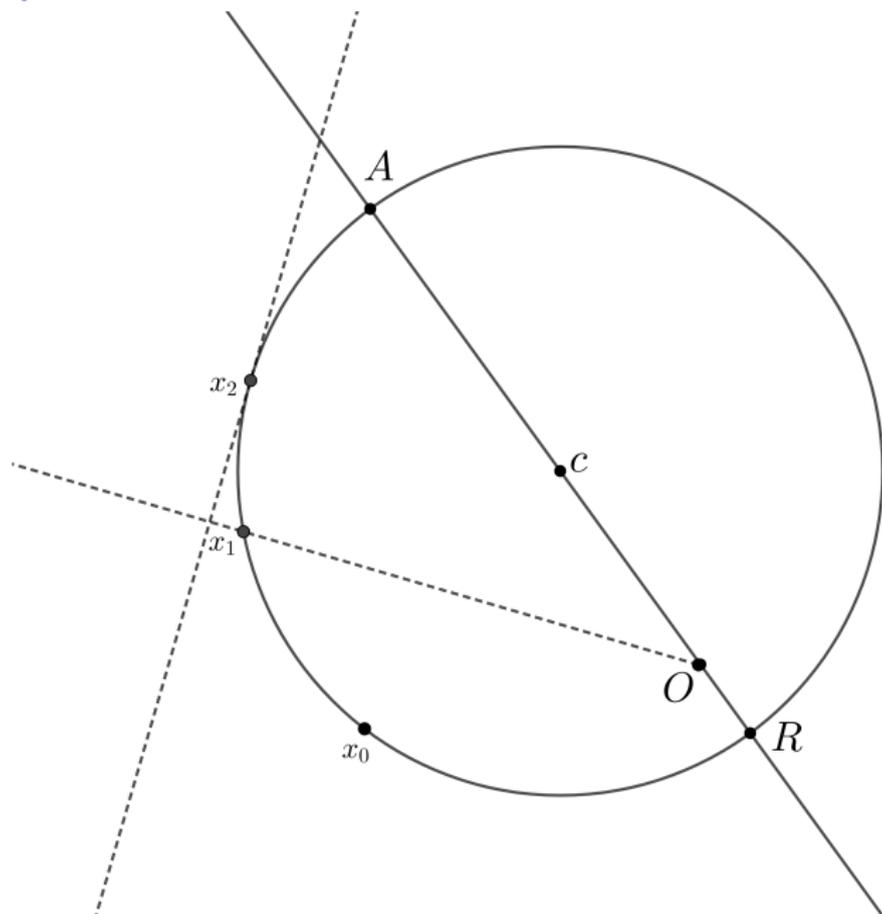
# Example 1



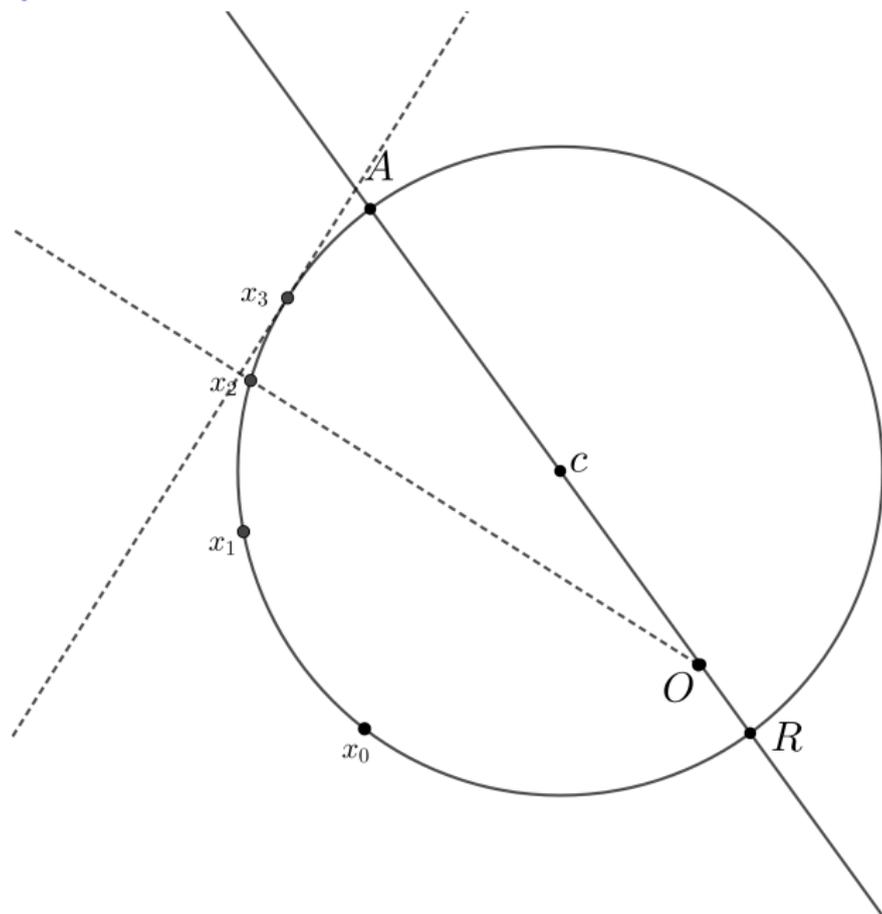
# Example 1



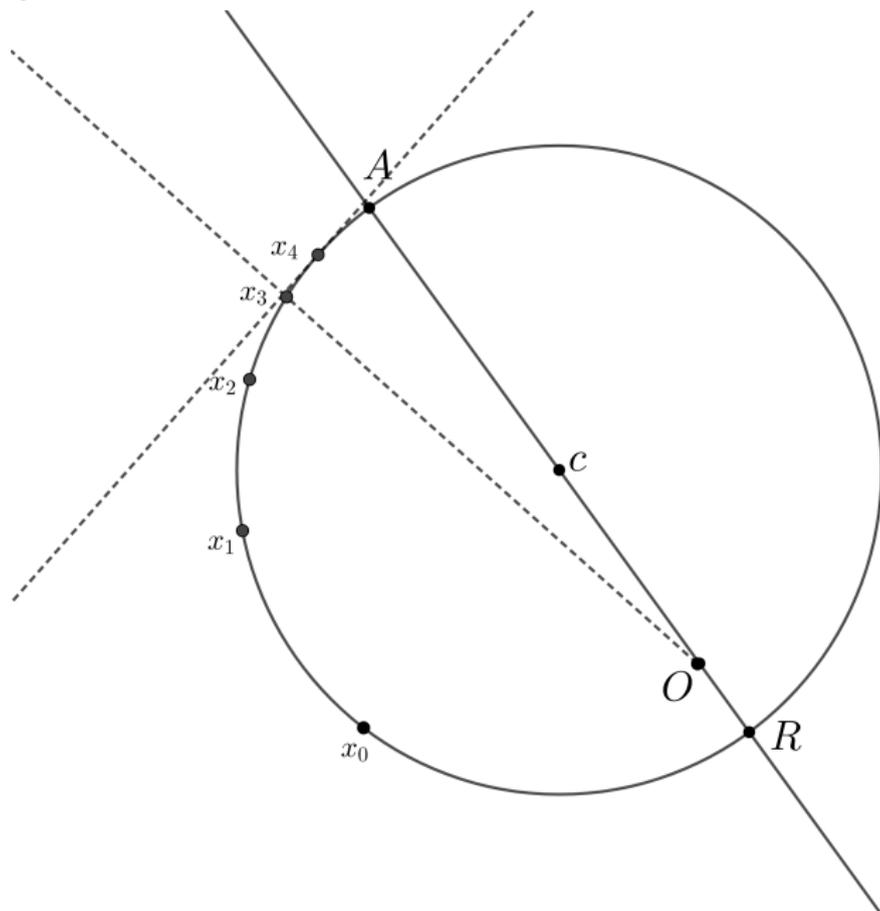
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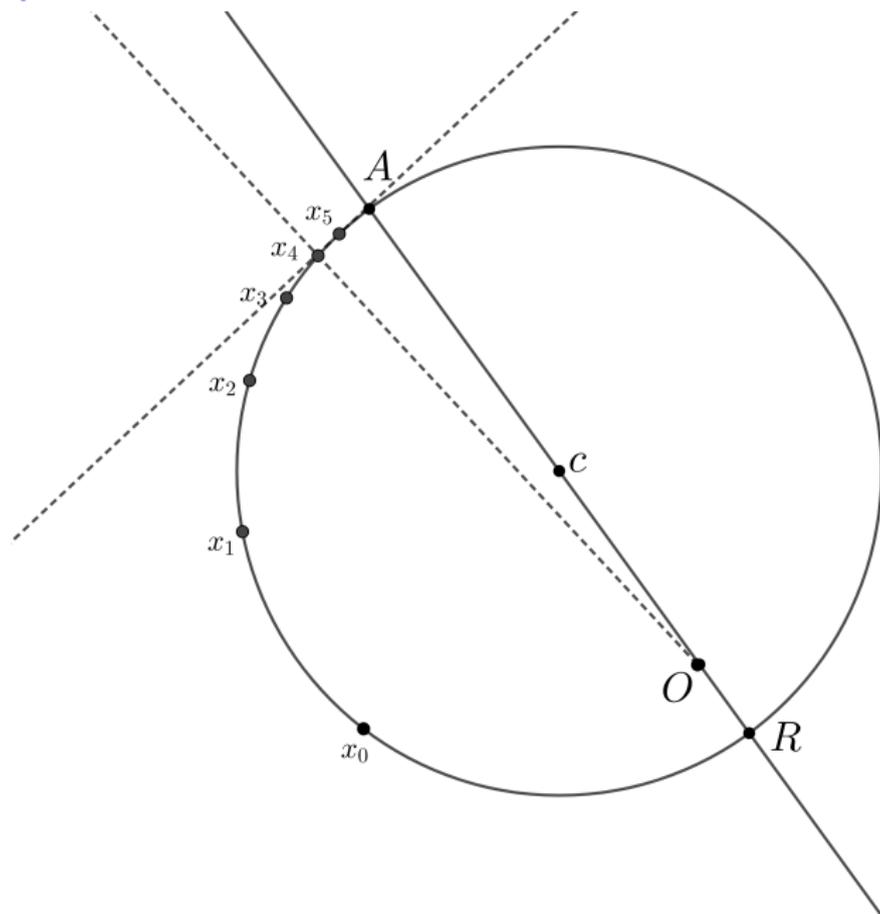
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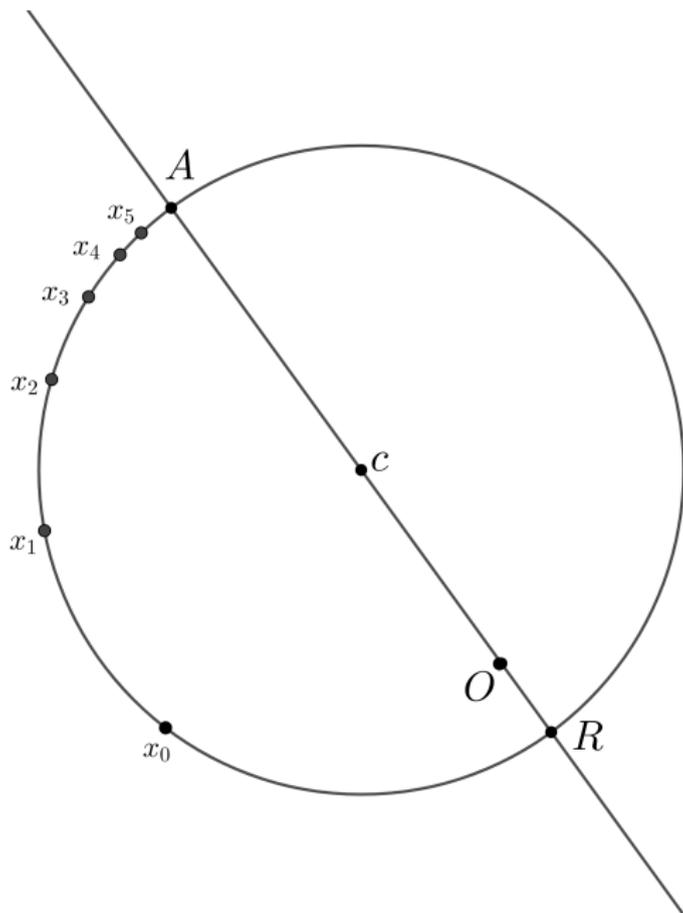
# Example 1



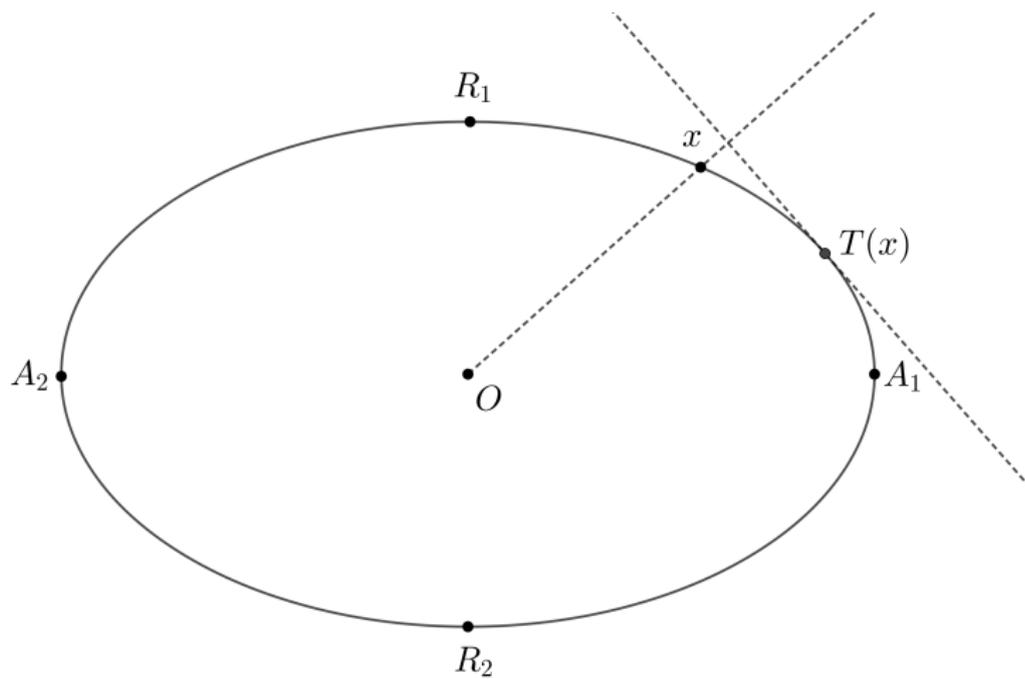
# Example 1



# Example 1



## Example 2



# Fixed point iteration

## Theorem (FKP20)

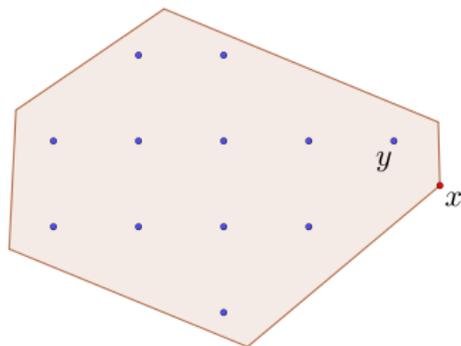
*Sequence  $\{x_0, x_1, x_2, \dots\}$  converges to a fixed point of  $T$ .*

Iterative method for maximizing  $\|x\|^2$  in  $\Delta$ .

Related to Franke-Wolfe optimization method.

# Convex relaxations

$$\operatorname{argmax}_{z \in S} f(z)$$



Combinatorial optimization via convex relaxation:

1. Discrete set of possible solutions  $S$  relaxed to convex set  $\Delta$ .
2. Optimize objective over  $\Delta$ .
3. “Round” solution  $x \in \Delta$  to solution  $y \in S$ .

Iteration with  $T$  can be used for rounding semidefinite relaxations.

# Elliptope

$\mathcal{S}_n = n$  by  $n$  symmetric matrices.

*Elliptope*  $\mathcal{L}_n \subseteq \mathcal{S}_n$  are positive semidefinite matrices with 1's on diagonal:

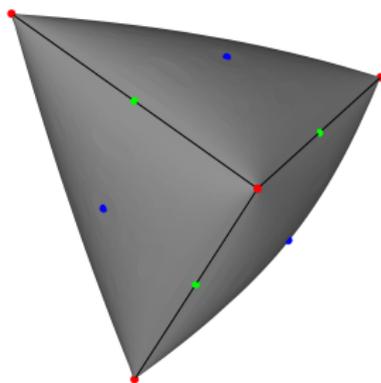
$$\mathcal{L}_n = \{X \in \mathcal{S}_n \mid X \succeq 0, X_{i,i} = 1\}.$$

- Goemans-Williamson semidefinite relaxation for max-cut.
- Gram matrices of  $n$  unit vectors in  $\mathbb{R}^n$ .
- Instance of *spectrahedron*.

## Fixed points in $\mathcal{L}_n$

$\mathcal{L}_3$  can be visualized in  $\mathbb{R}^3$ .

$$X = \begin{pmatrix} 1 & x & y \\ x & 1 & z \\ y & z & 1 \end{pmatrix}$$



Red fixed points are irreducible matrices with rank 1.  
Blue fixed points are irreducible matrices with rank 2.  
Green fixed points are reducible matrices with rank 2.

# Algebraic Characterization

## Lemma (FKP20)

*If  $X = T(M)$  there exists a diagonal matrix  $D$  such that*

$$MX = DX.$$

*(similar to eigenvector  $Mx = \lambda x$ )*

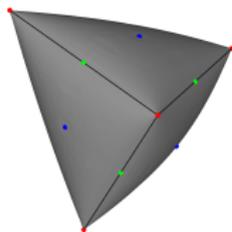
## Theorem (FKP20)

*$X = T(X)$  iff there exists a diagonal matrix  $D$  such that*

$$X^2 = DX.$$

# Elliptopes

$\mathcal{L}_3$ : finite number of fixed points.



$\mathcal{L}_4$ : infinite number of fixed points.

Any  $-1 < c < 1$  leads to a distinct fixed point:

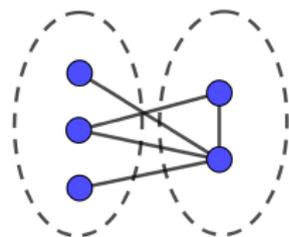
$$X(c) = \begin{pmatrix} 1 & -\sqrt{1-c^2} & 0 & c \\ -\sqrt{1-c^2} & 1 & -c & 0 \\ 0 & -c & 1 & -\sqrt{1-c^2} \\ c & 0 & -\sqrt{1-c^2} & 1 \end{pmatrix}$$

$\mathcal{L}_n$ : finite number of *regular* fixed points.

(one-dimensional normal cone)

# Rounding max-cut relaxation

Max-cut:  
partition vertices of graph maximizing  
the weight of edges between sets.



Semidefinite relaxation involves linear optimization over  $\mathcal{L}_n$ .

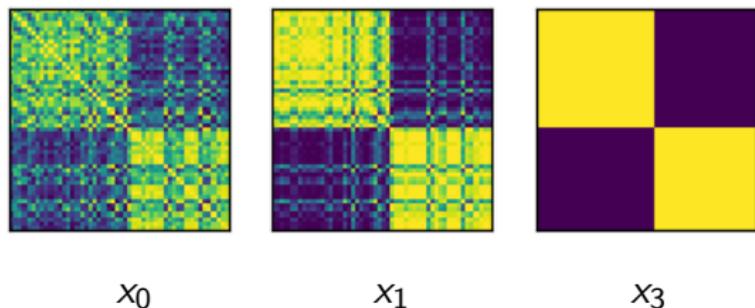
Partitions of  $\{1, \dots, n\}$  are the vertices of  $\mathcal{L}_n$ .

Round  $X \in \mathcal{L}_n$  by finding the closest vertex  $Y$ .

- Relax to  $\mathcal{L}_n$ :  $Y=T(X)$ .
- If  $Y$  is *not* a vertex, we iterate to find vertex close to  $Y$ .

The vertices of  $\mathcal{L}_n$  are the attractive fixed points of  $T$ .

## Rounding max-cut relaxation



Fixed point iteration starting from the solution of the max-cut relaxation for a graph with 50 vertices and random weights.