

# Bounded Integer Linear Constraint Solving via Lattice Search

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# Overview

2

- We present an **algorithm** for solving systems of **integer linear constraints** where each linear form has **lower** and **upper bounds**.
  - We call these **bounded** integer linear problems.
- We implemented this in a tool called **BLT**.
- BLT has proven many orders of magnitude faster on a class of **preimage** problems arising from **JPEG**.

# Overview

- We present an **algorithm** for solving systems of **integer linear constraints** where each linear form has **lower** and **upper bounds**.
- We call these **bounded** integer linear problems.
- We implemented this in a tool called **BLT**.

- **BLT has preimage problems**

On one of the “easier” problems, we ran several SMT solvers for months without success.

BLT takes under a second to find a solution.

# Bounded Integer Linear Constraints

4

- **Bounded** integer linear constraint problems have the form below:

$$\begin{aligned} & l_1 \leq a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n \leq u_1 \\ \wedge & l_2 \leq a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n \leq u_2 \\ & \dots \\ \wedge & l_m \leq a_{m1}x_1 + a_{m2}x_2 + \cdots + a_{mn}x_n \leq u_m \end{aligned}$$

- Both the lower and upper bounds  $l_i$  and  $u_i$  coefficients  $a_{ij}$  are rational constants.
- The free variables  $x_j$  are all integers.

# Bounded Integer Linear Constraints

5

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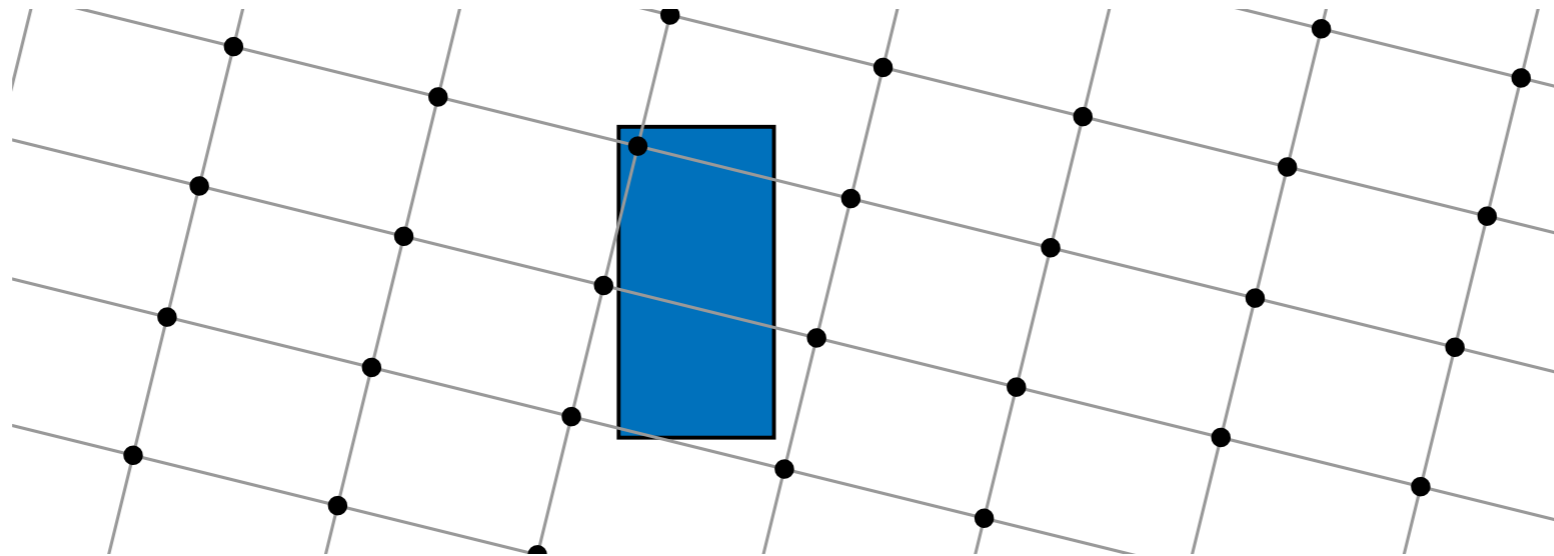
$$l \leq Ax \leq u$$

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- The free variables  $x_j$  are all integers.

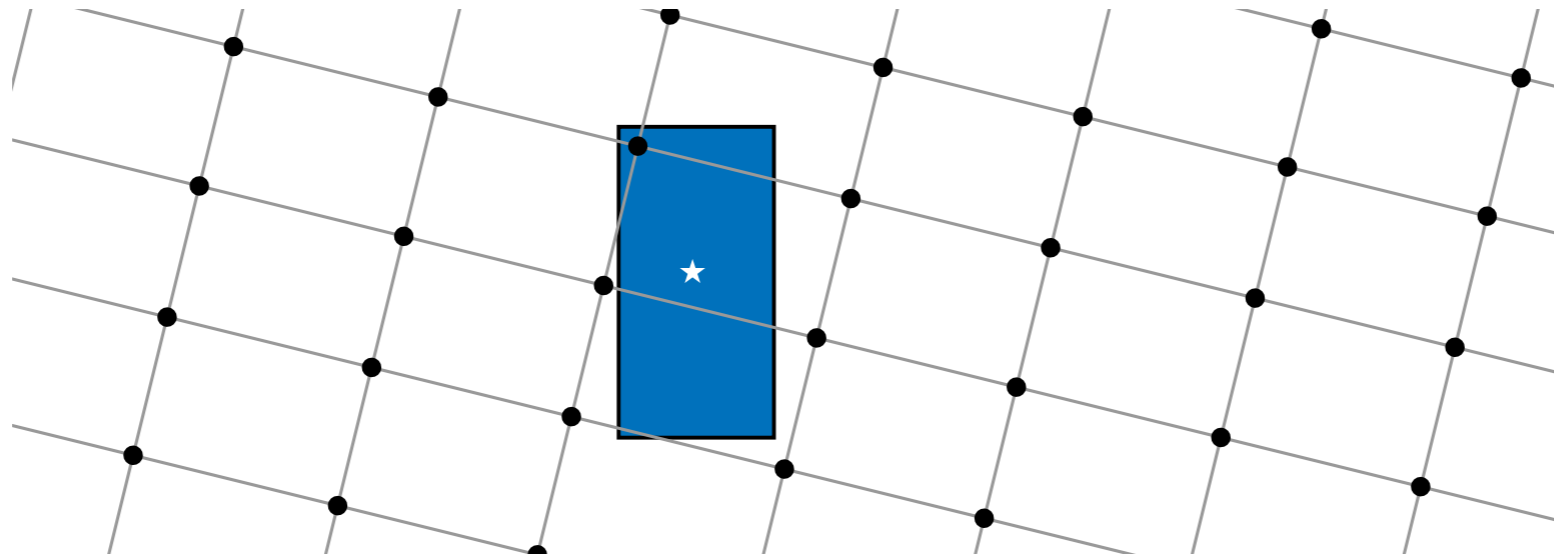
- The two problems below are equivalent:
  - Finding a solution to  $l \leq Ax \leq u$ .
  - Finding a point in the intersection of the lattice generated by the columns of  $A$  and hyper-rectangle with corners  $l$  and  $u$ .

$$l \leq A_0x_0 + A_1x_1 + \dots + A_nx_n \leq u$$

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- Intuitively, it makes sense to look for points near the center of the rectangle.



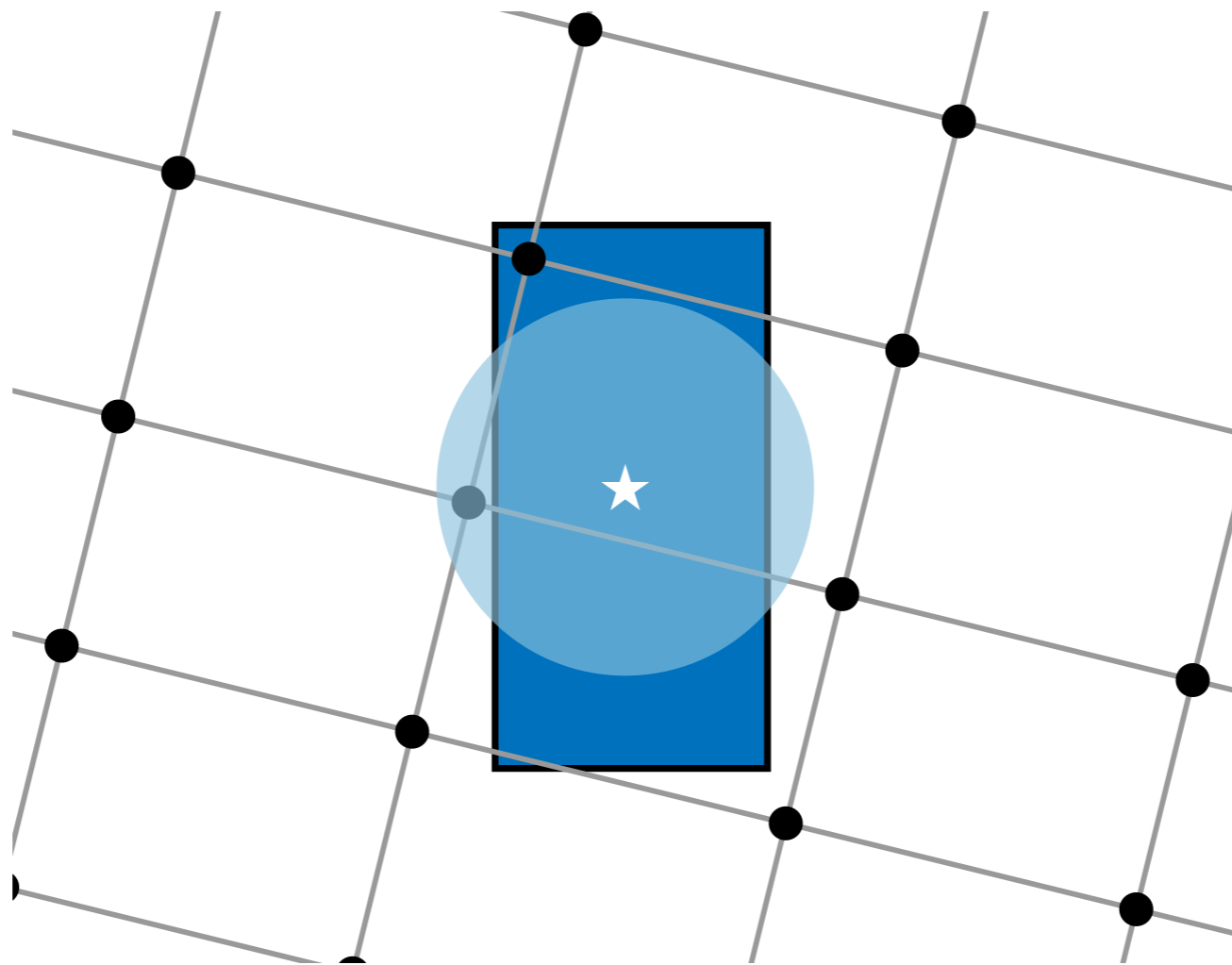
# Closest Vector Problem

9

- Is there an algorithm to find a lattice point close to a given location?
  - Yes! Schnorr-Euchner is a well known exponential-time algorithm for this problem.
  - It incrementally builds up a partial assignment making guesses using a **nearest hyperplane** heuristic.
  - It detects when there are no real solutions within a given distance of the target point and backtracks.
- However, it uses the  $L^2$  distance—not what we want.

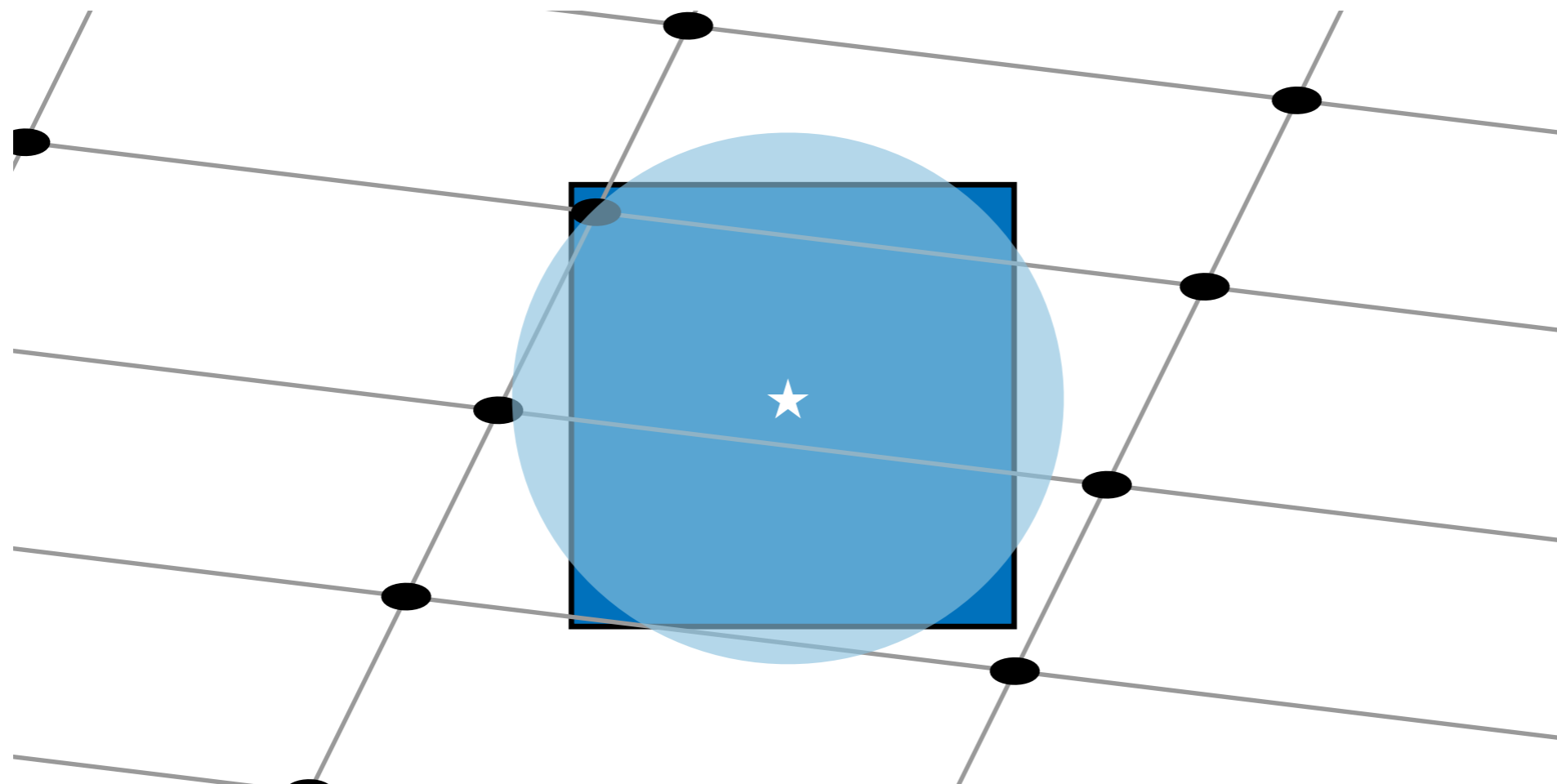
# $L^2$ Approximation

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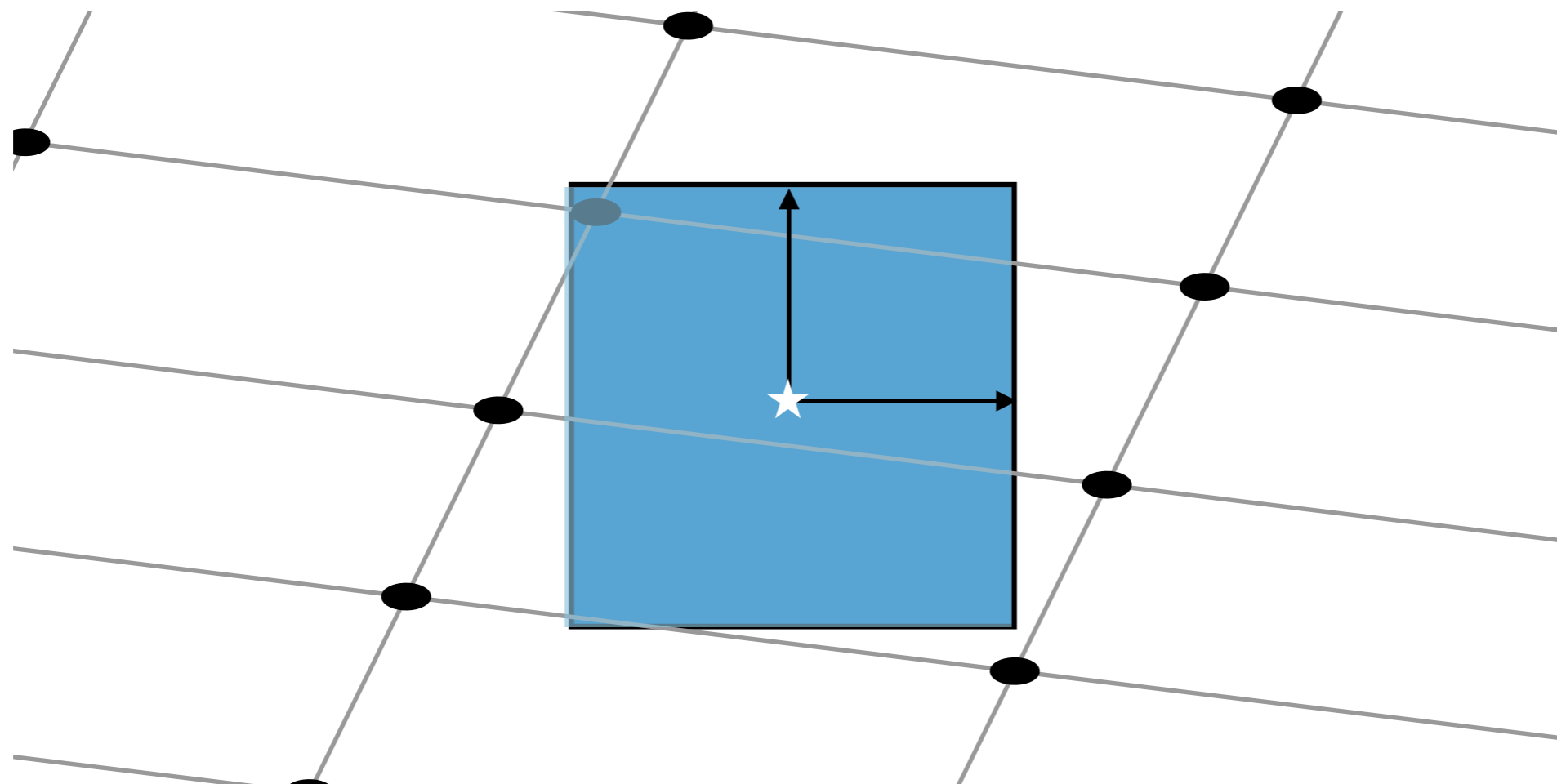
# Normalize Dimensions

11



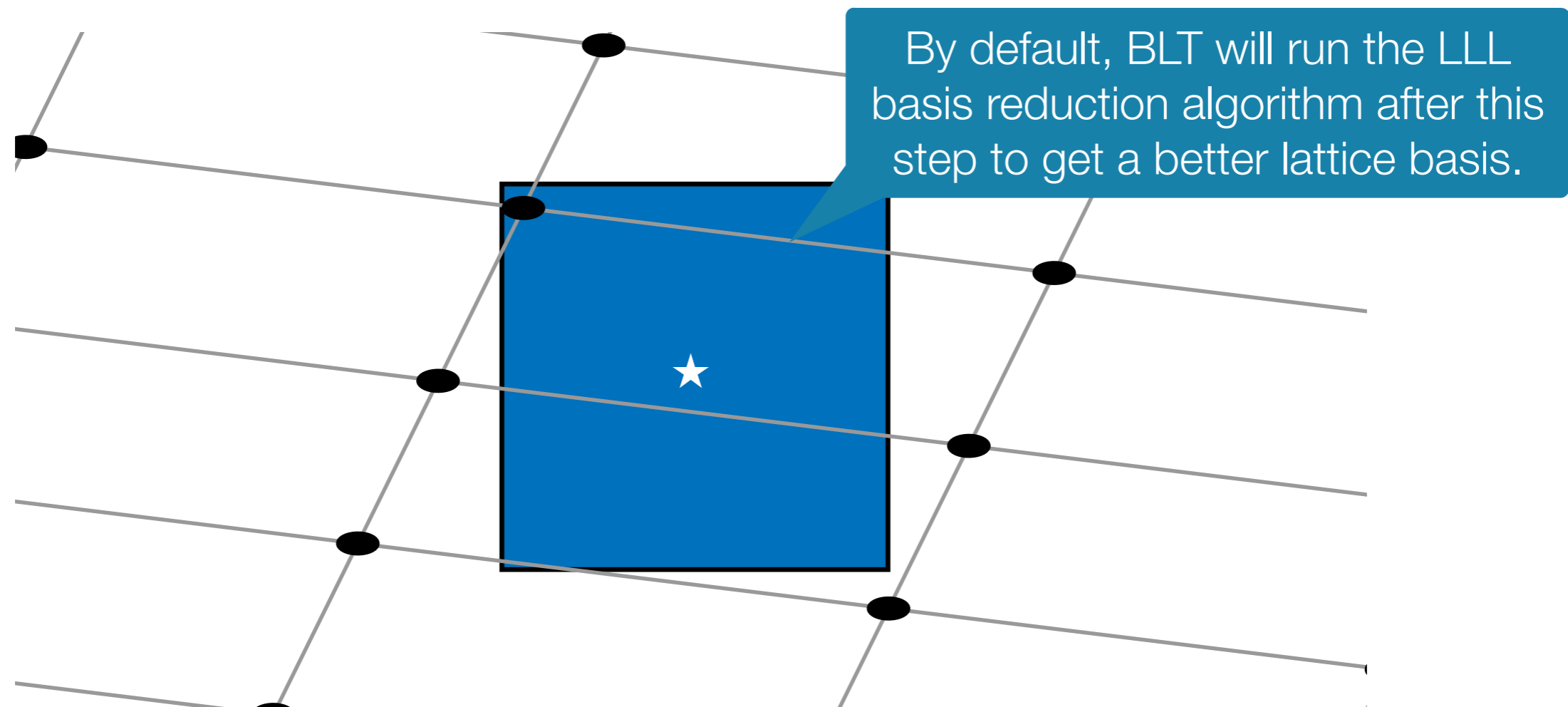
# Use $L^\infty$ Distance

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# Normalize Dimensions

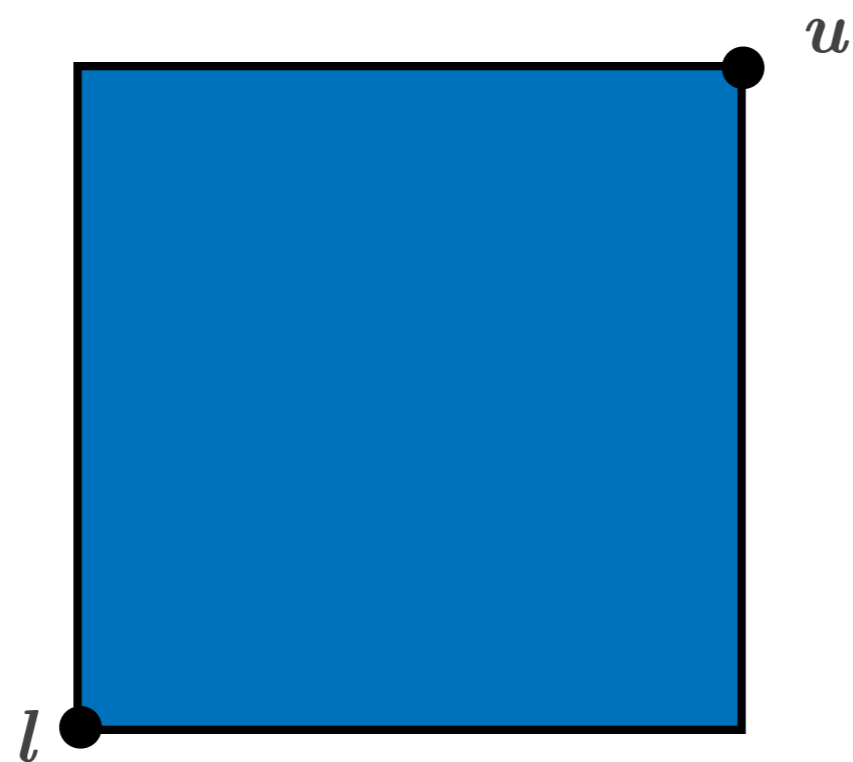
13



# The Algorithm

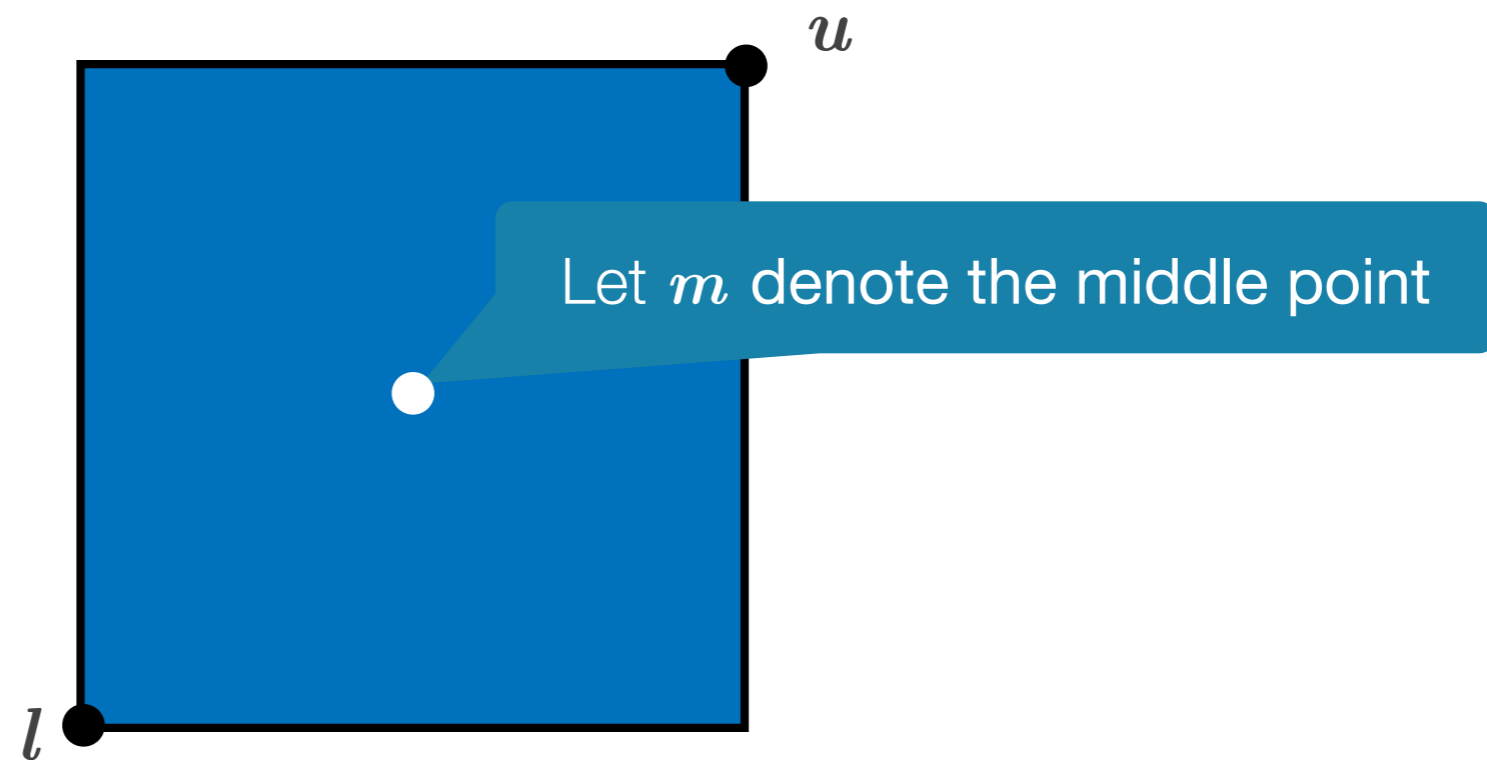
# The Hypercube

15



# The Hypercube

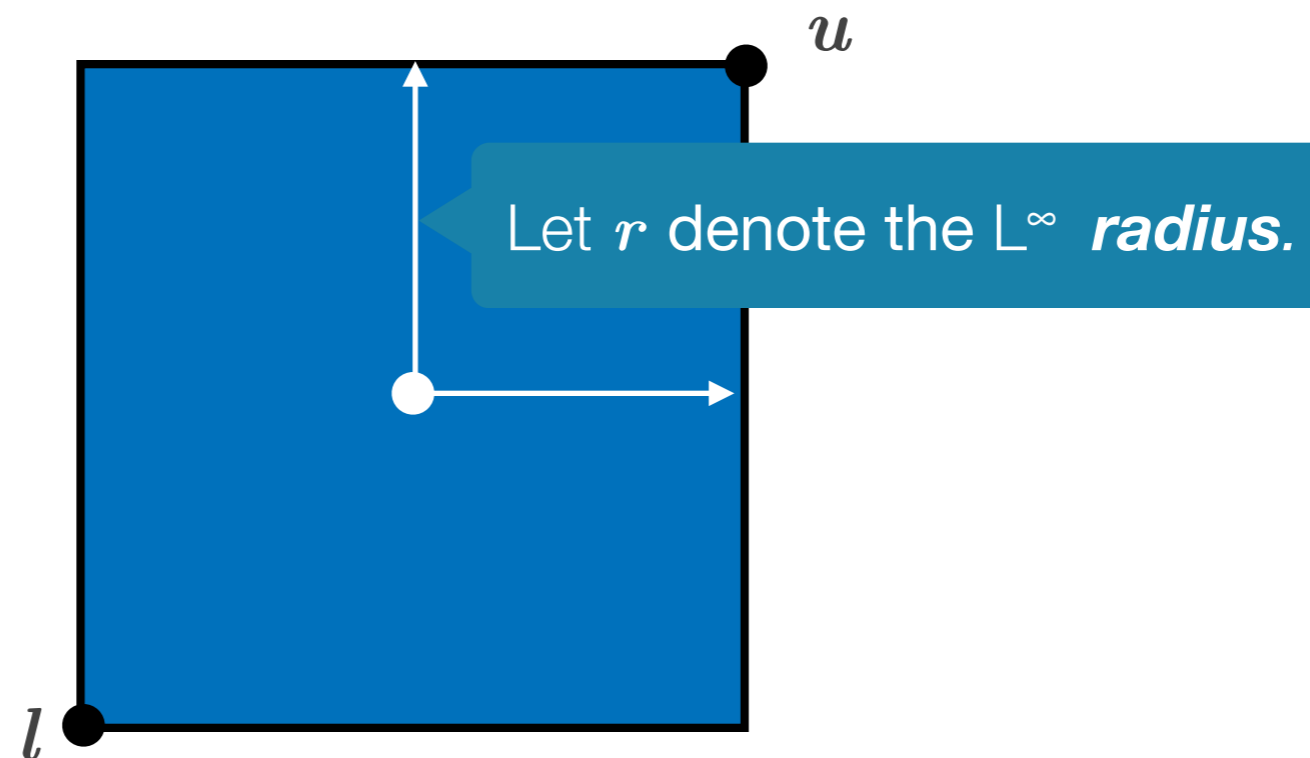
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# The Hypercube

17



# $L^\infty$ Distance

18

- The  $L^\infty$  distance between two points  $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$  is just the largest difference between two coefficients.

$$d_\infty(\mathbf{x}, \mathbf{y}) := \max\{|x_i - y_i| \mid i = 1, \dots, n\}$$

- This also extends to subsets  $X, Y \subseteq \mathbb{R}^n$

$$d_\infty(X, Y) := \min_{(\mathbf{x}, \mathbf{y}) \in X \times Y} d_\infty(\mathbf{x}, \mathbf{y})$$

- When  $X$  and  $Y$  can be described by a system of linear inequalities, then the distance can be solved by linear programming.

$$\begin{aligned}d_{\infty}(X, Y) &= \min \{ d_{\infty}(x, y) \mid x \in X, y \in Y \} \\ &= \min \left\{ \max_i \{ |x_i - y_i| \} \mid x \in X, y \in Y \right\} \\ &= \min \{ t \mid |x_i - y_i| \leq t, x \in X, y \in Y \} \\ &= \min \{ t \mid x_i - y_i \leq t, -(x_i - y_i) \leq t, x \in X, y \in Y \}\end{aligned}$$

# Lattice Definitions

20

- A **Lattice** over a fixed basis  $\{\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n\}$  is the set of points that can be expressed as an integral sum of the vectors in the basis.

$$\mathcal{L} = \left\{ \sum_{i=1}^n c_i \mathbf{a}_i \mid c_i \in \mathbb{Z} \right\}$$

- Given a partial mapping  $\theta$  from variables to integers, we can define the **sublayer** of  $\theta$  as follows:

$$\mathcal{L}_\theta^{\mathbb{Z}} := \left\{ \sum_{i=1}^n c_i \mathbf{a}_i \mid c_i \in \mathbb{Z}, i \in \text{dom}(\theta) \Rightarrow c_i = \theta(i) \right\}$$

# Lattice Definitions

21

- To prune solutions, we use the real affine linear space below:

$$\mathcal{L}_\theta^{\mathbb{R}} := \left\{ \sum_{i=1}^n c_i \mathbf{a}_i \mid c_i \in \mathbb{R}, i \in \text{dom}(\theta) \Rightarrow c_i = \theta(i) \right\}$$

# BLT Algorithm

22

- Our search procedure maintains a partial assignment  $\theta$  from variables to integers with the invariant:

$$d_{\infty}(\mathcal{L}_{\theta}^{\mathbb{R}}, m) \leq r$$

- We start with the empty assignment.
- At each step we first choose an unbound variable  $y$  to add to the substitution.

- BLT chooses the assignment to  $y$  by finding the value  $c \in \mathbb{R}$  that minimizes the distance:

$$d_{\infty}(\mathcal{L}_{\theta}^{\mathbb{R}}[y \mapsto c], m)$$

- The first assignment to  $y$  is chosen by rounding  $c$  to the nearest integer
- If we need to backtrack, we incrementally explore values starting from those near  $c$  and working outwards

$$\dots, \lfloor c - 1 \rfloor, \lfloor c \rfloor, \lceil c \rceil, \lceil c + 1 \rceil, \dots$$

- We show in the paper that the number of values is finite, and thus our procedure will eventually terminate.

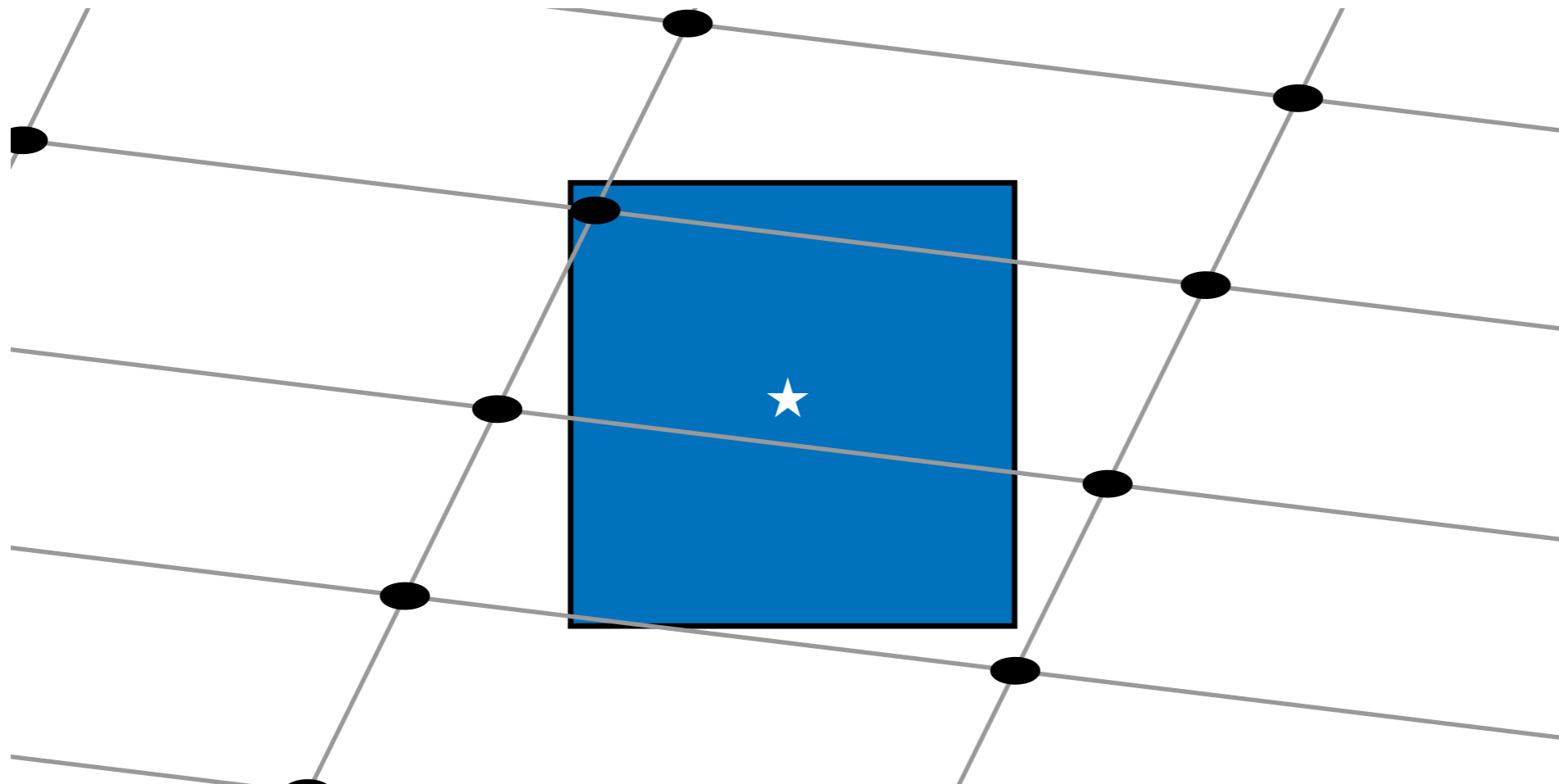
Example

24



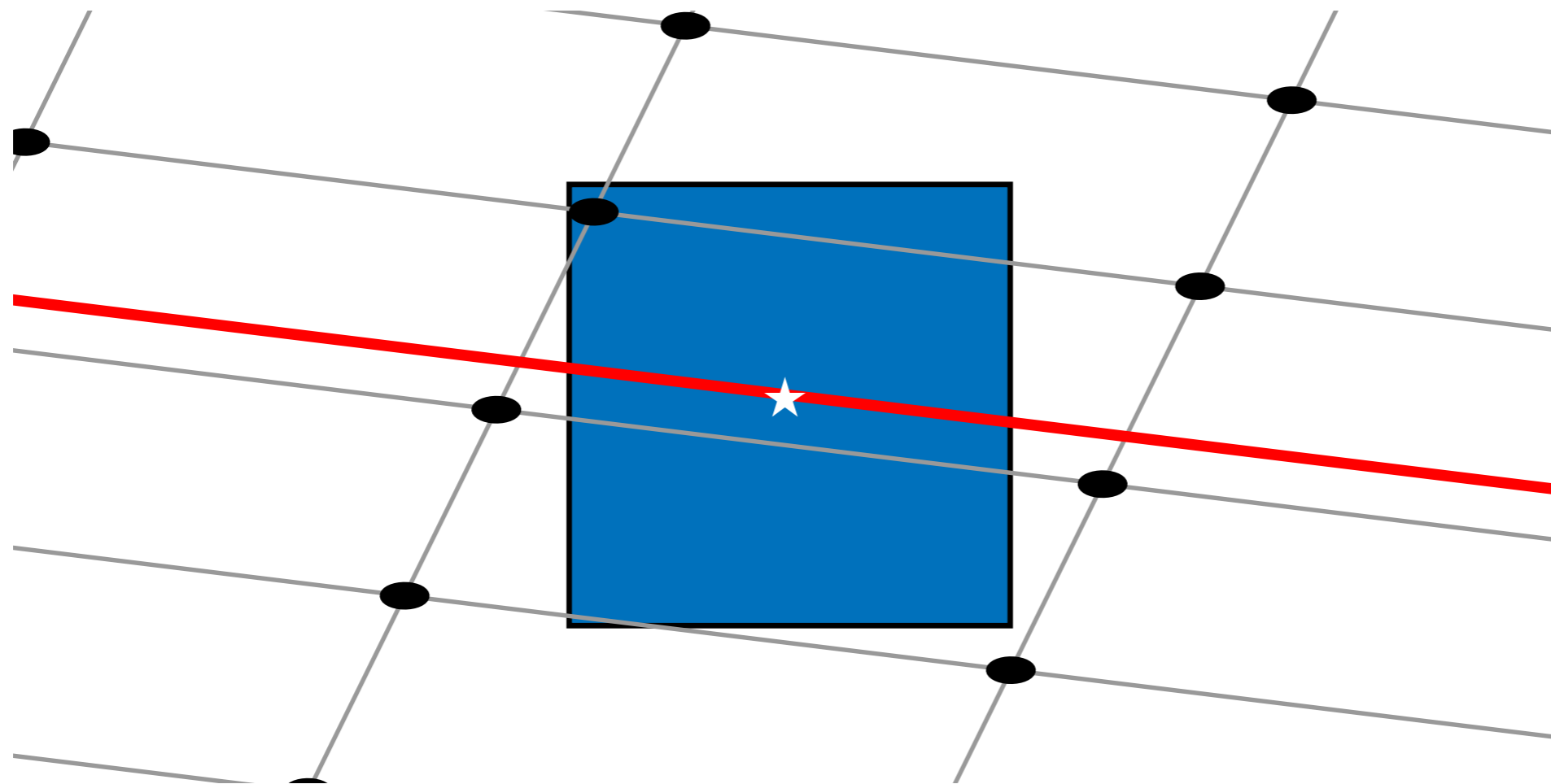
# Example

25



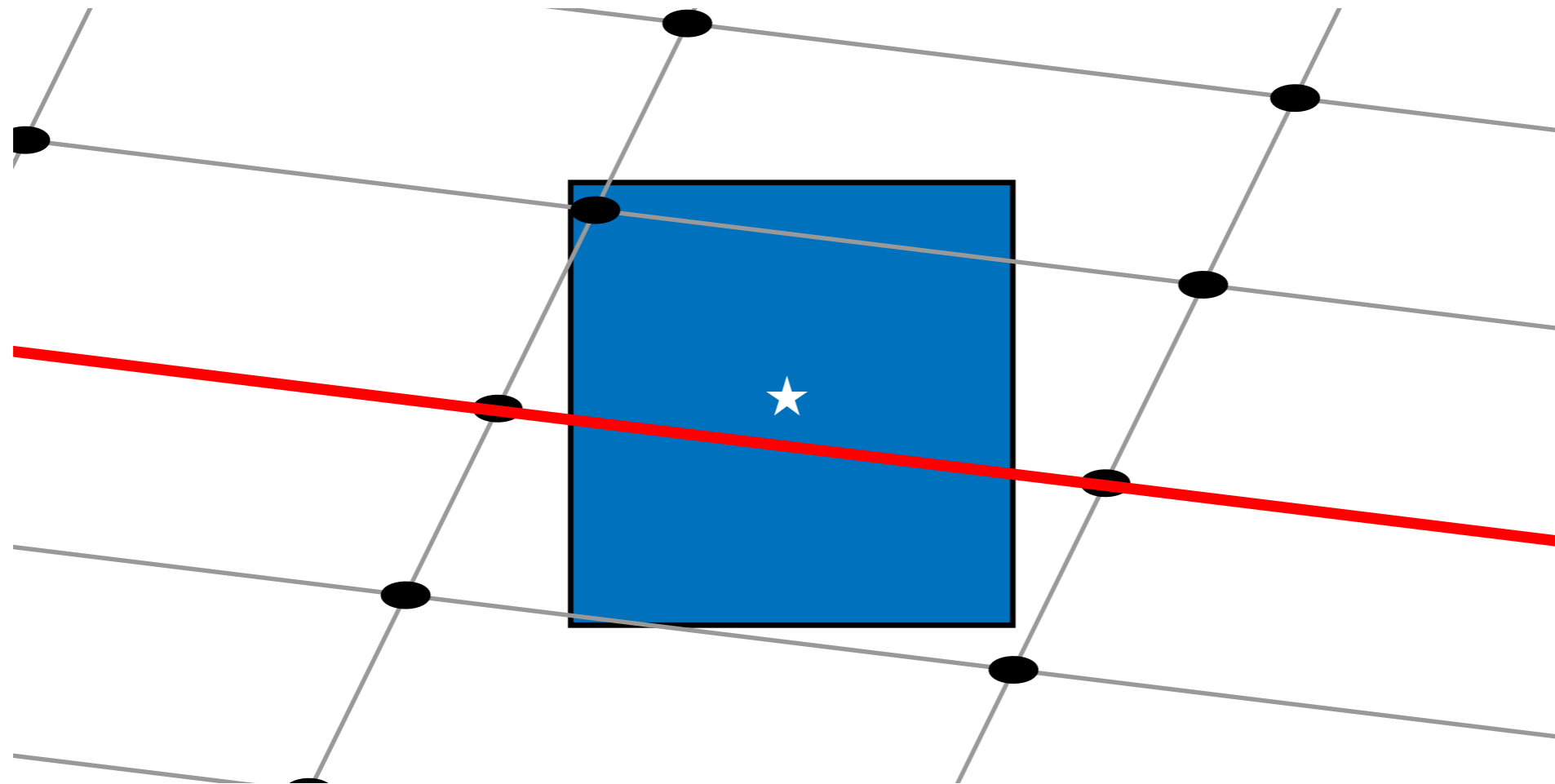
# First Choice

26



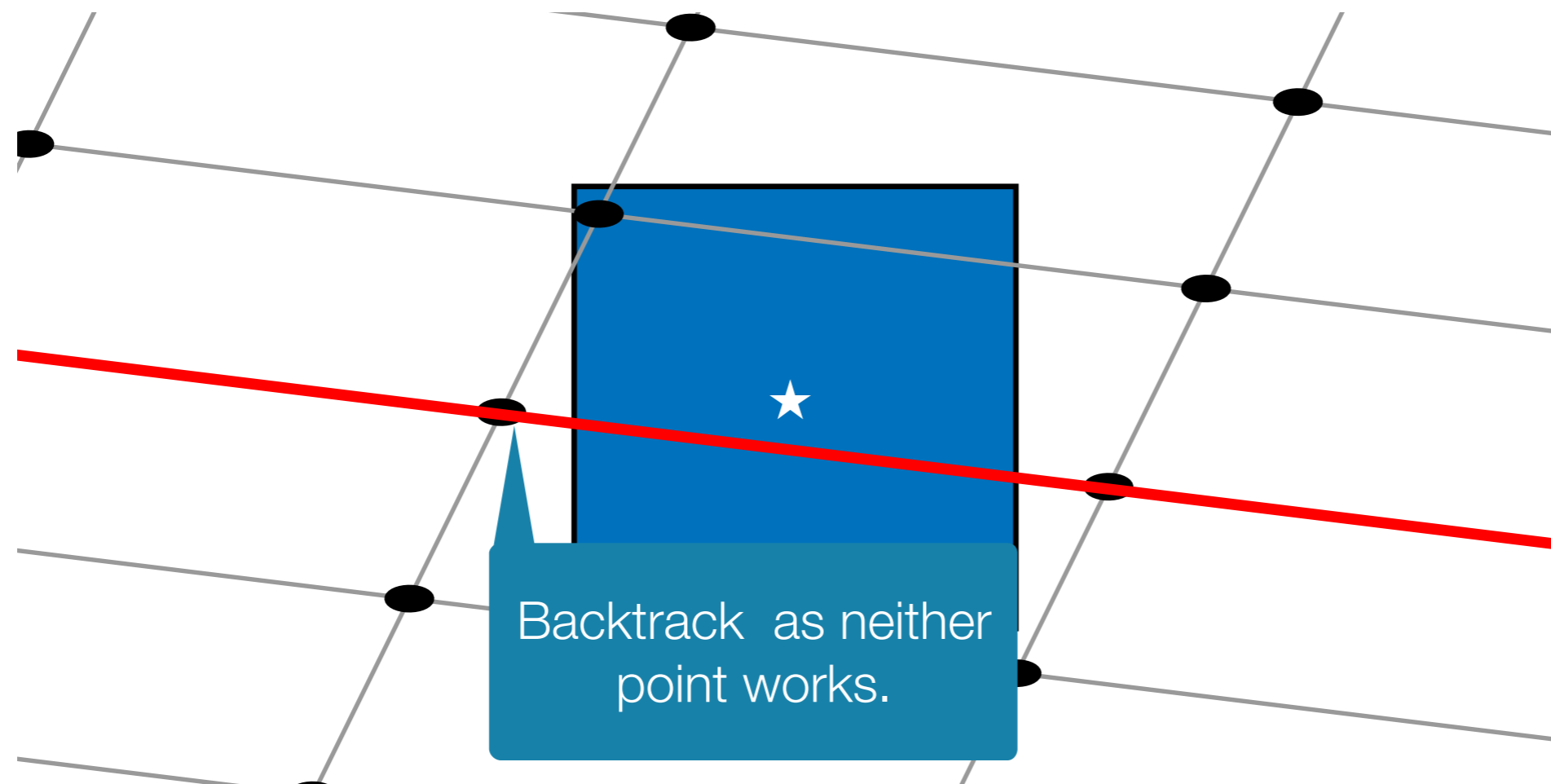
# First Choice

27



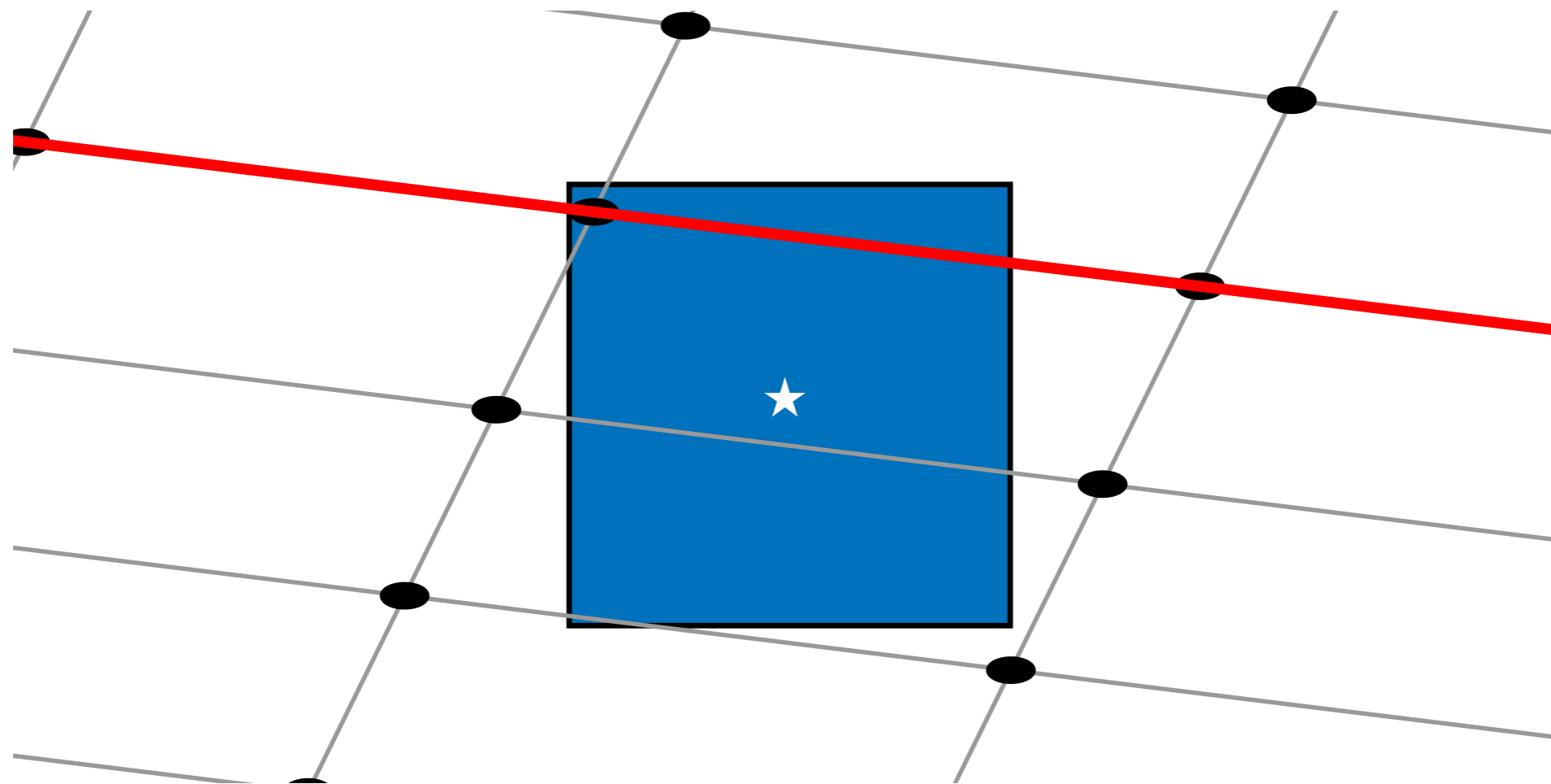
# First Choice

28



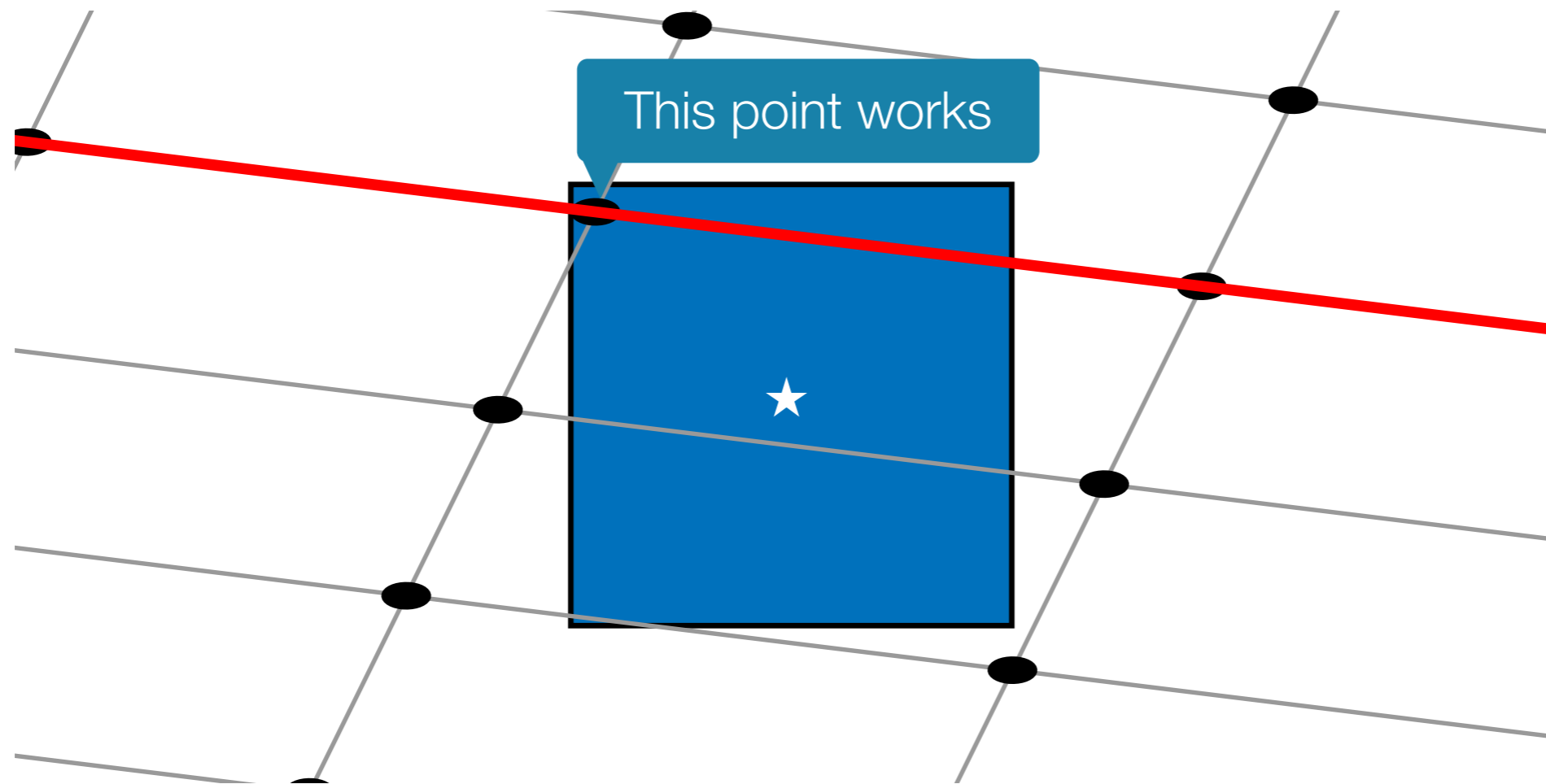
# Second Choice

29



# Second Choice

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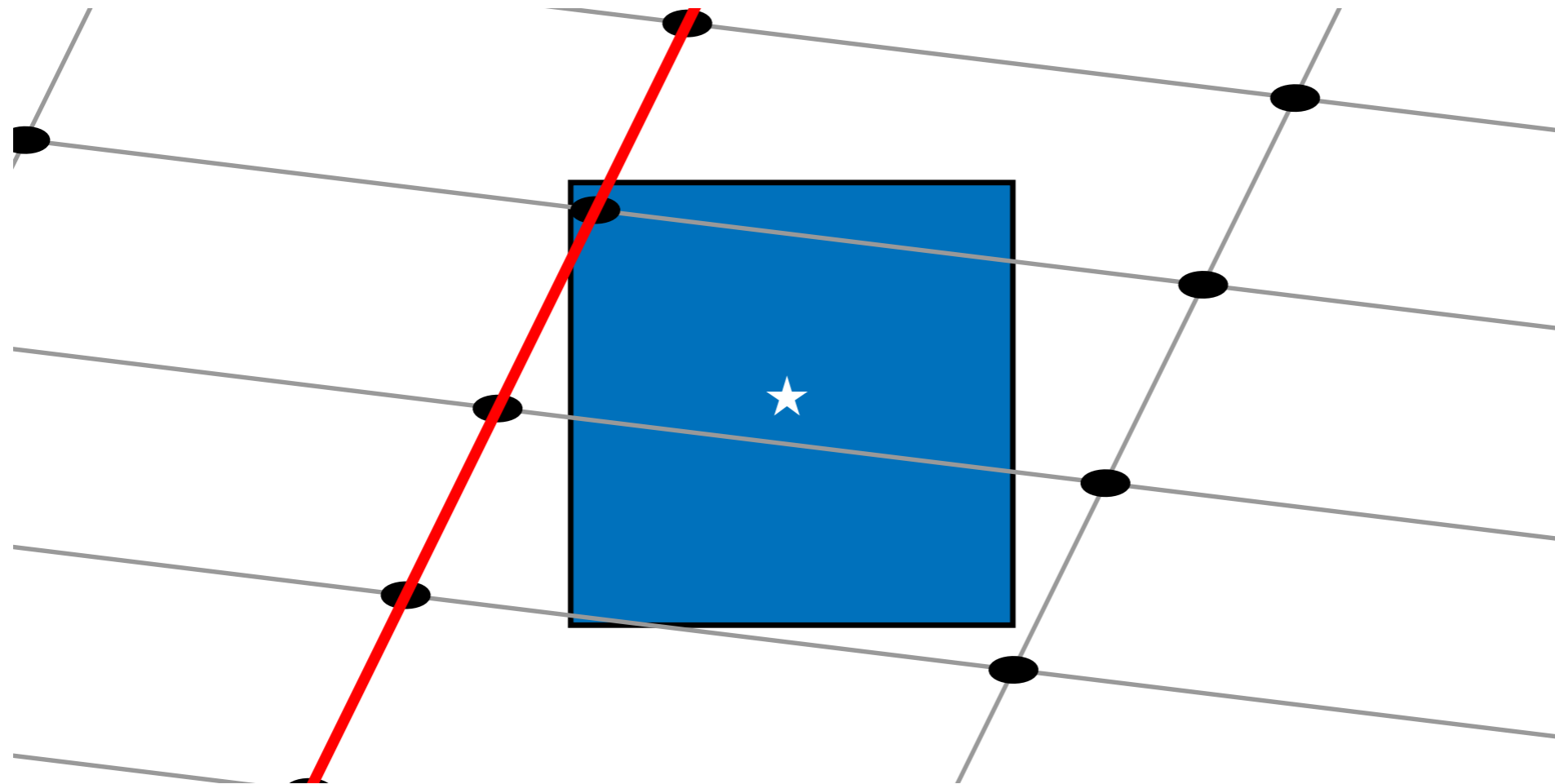
# Longest Vector First

31

- This approach worked, but required backtracking.
- As an additional heuristic, BLT will always assign values to variables in the order of decreasing size of the associated basis vector.

# Better assignment

32





# JPEG Preimages

# JPEG Preimages

34

- It turns out that JPEG decompression can be expressed as linear function from integer coefficients in a file.
- Each 8x8 block of pixels in a single channel is computed from 64 integer coefficients  $\mathbf{C}$  that are stored in a compressed form

$$\mathbf{I} = \text{idct}_2(\mathbf{Q}_{lvl} * \mathbf{C}) + 128$$

# JPEG Preimages

35

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- Each 8x8 block of pixels in a single channel is computed from 64 integer coefficients  $\mathbf{C}$  that are stored in a compressed form

$$\mathbf{I} = \text{idct}_2(\mathbf{Q}_{lvl} \cdot * \mathbf{C}) + 128$$

The coefficients in  $\mathbf{Q}_{lvl}$  depend on the JPEG quality level.

High Compression Ratio = Low Quality Level = Large values in  $\mathbf{Q}_{lvl}$

# JPEG Preimages

36

- In the paper, we pose the problem of finding a set of compressed coefficients so that the decompressed image satisfies precise constraints on specific pixels.

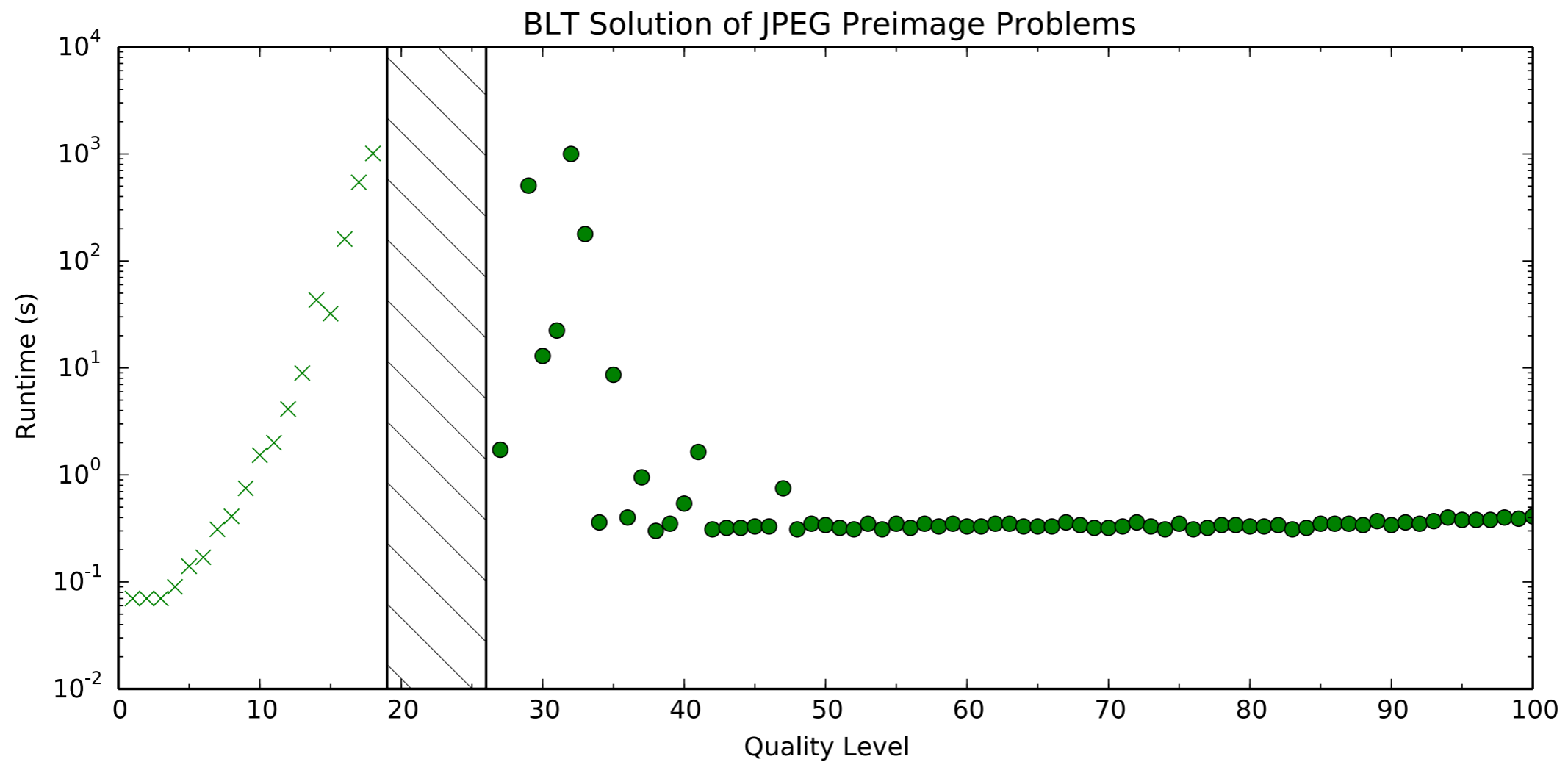
Decompression  
 $C \longrightarrow O$

		h	e	l	l	o	
		w	o	r	l	d	!

- We show how this can in turn be converted into a bounded ILP problem and solved efficiently using BLT.

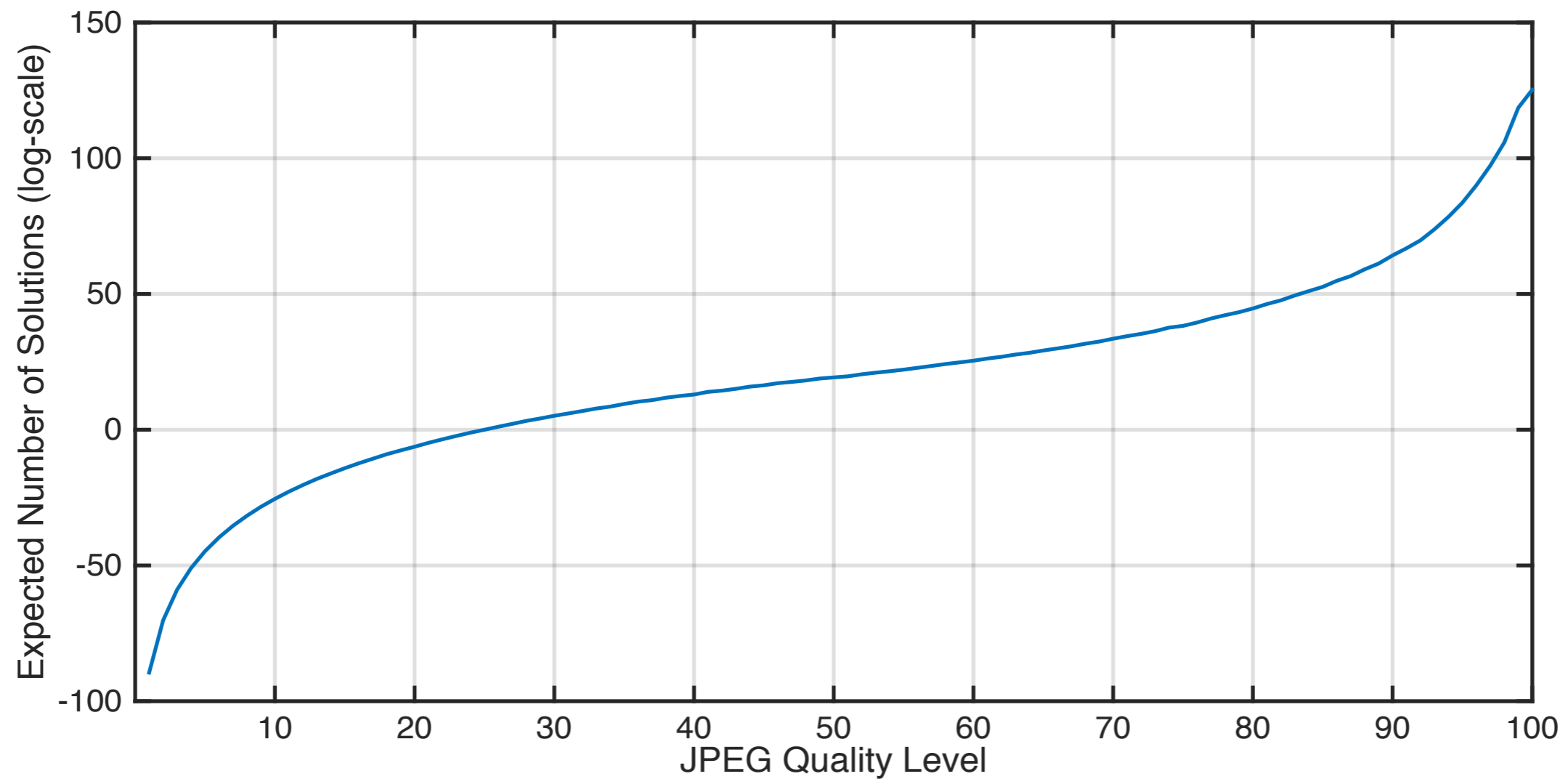
# BLT Benchmark Results

37



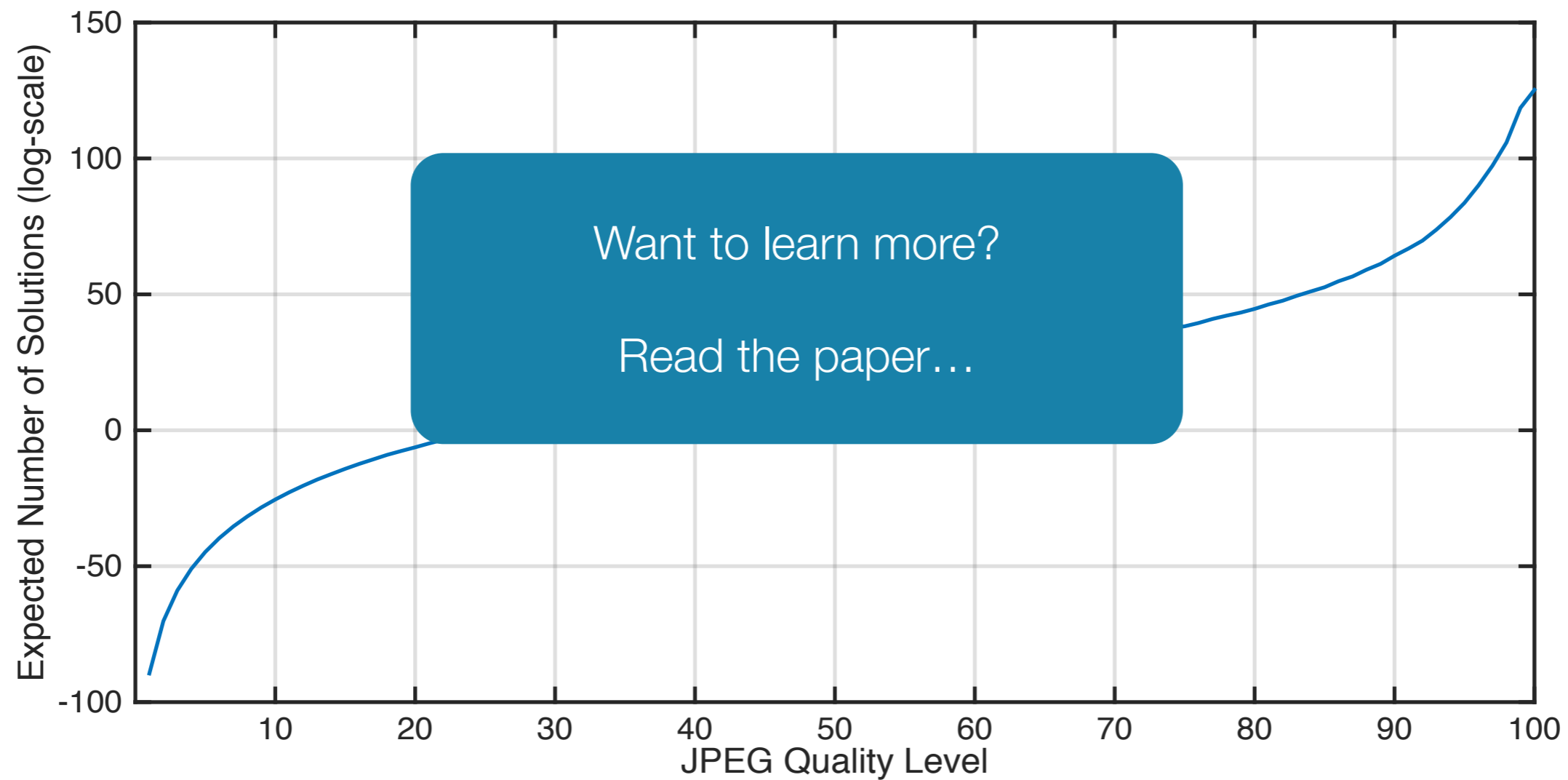
# Number of Solutions

38



# BLT Benchmark Results

39



# Implementation Experience

40

- The algorithm is implemented in a C++ library called BLT.
  - BLT uses GLPK for solving real linear problems.
  - GLPK uses floating point; UNSAT results may be incorrect.
- In addition to CVC4, Yices, Z3, we have also tried:
  - Bit-level version of the problem using Boolector, abc, and glucose.
  - Optimization tools such as Gurobi and GLPK.



# Conclusions

41

- This paper describes an algorithm for solving ILP problems with the form

$$l \leq Ax \leq u$$

- We have shown the 2-dimensional case, but the algorithm works for arbitrary systems of linear equations with this form.
- We have not seen Schnorr-Euchner generalized in this way, nor have we seen it used for solving linear equations.
- In future work, we would like to explore ways to integrate this into SMT solvers.
- Once we have finished some remaining cleanups, we plan to release BLT and SMTLIB 2 encodings of the JPEG problem.

# Thanks

We'd like to thank David Flamm, Patrick Lincoln, Dejan Jovanović, and Grant Passmore for feedback and discussions that helped develop the ideas in this paper.