

# Encodings in Mixed Integer Linear Programming

Juan Pablo Vielma

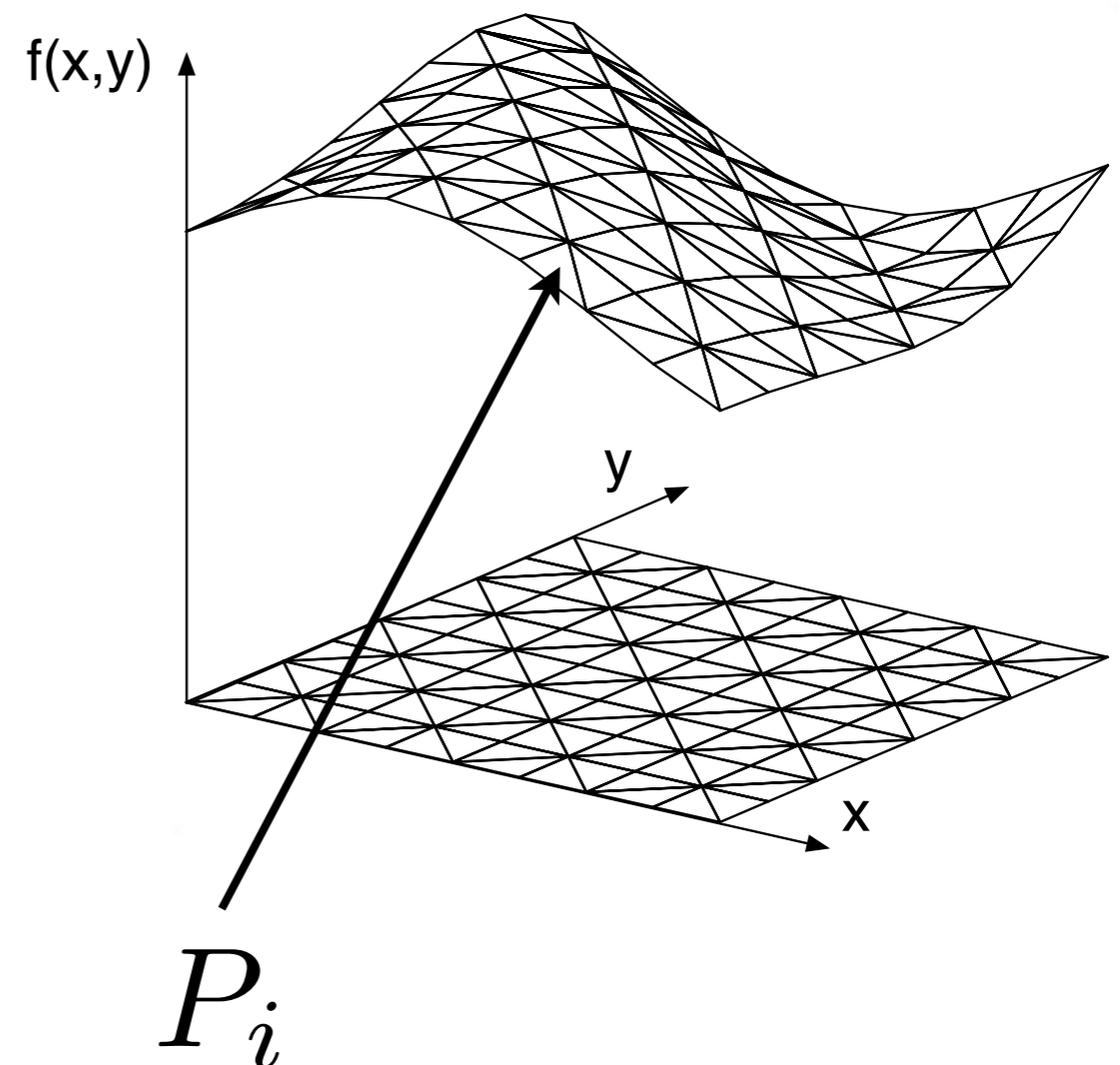
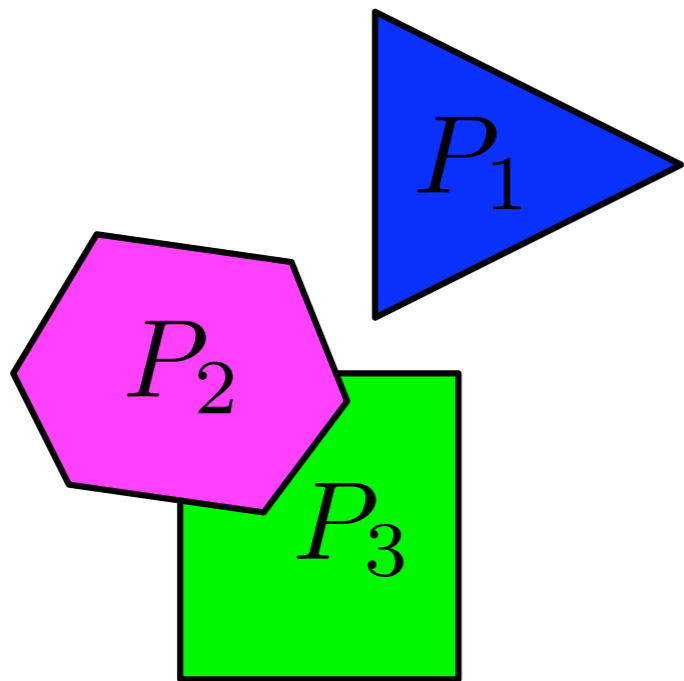
*Sloan School of Business,  
Massachusetts Institute of Technology*

Universidad de Chile,  
December, 2013 – Santiago, Chile.

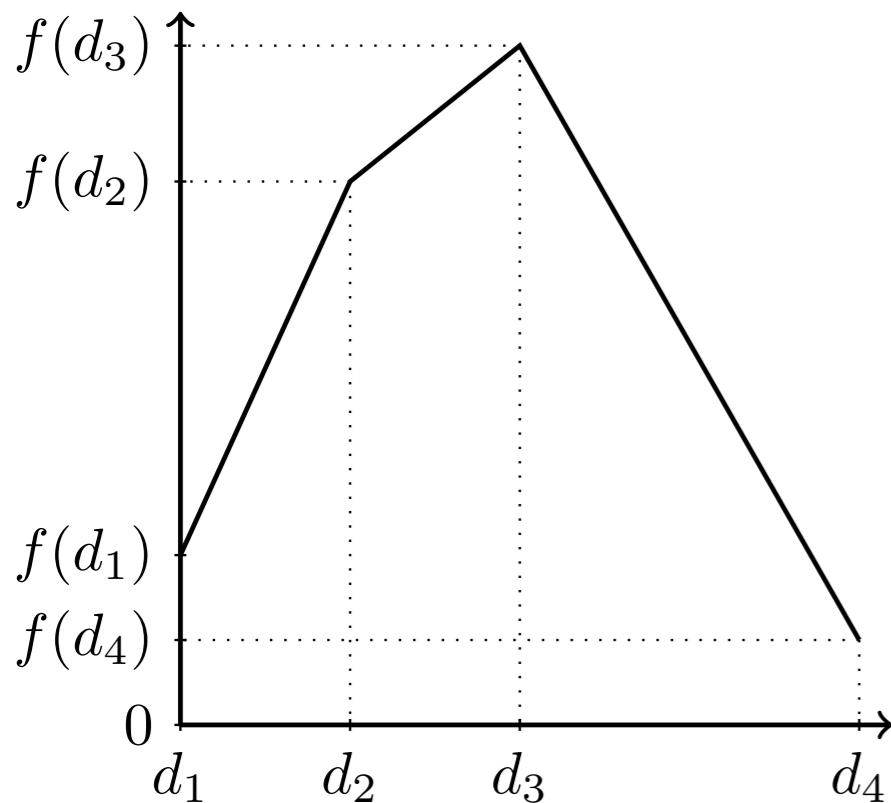
# Mixed Integer Binary Formulations

- MIP Formulations = Model Finite Alternatives

$$x \in \bigcup_{i=1}^n P_i \subseteq \mathbb{R}^d$$



# Textbook Formulation



Formulation for  $f(x)=z$

$$\sum_{i=1}^4 d_i \lambda_i = x,$$

$$\sum_{i=1}^4 f(d_i) \lambda_i = z$$

$$\sum_{i=1}^4 \lambda_i = 1,$$

$$\lambda_i \geq 0$$

$$\sum_{i=1}^3 y_i = 1,$$

$$y_i \in \{0, 1\}$$

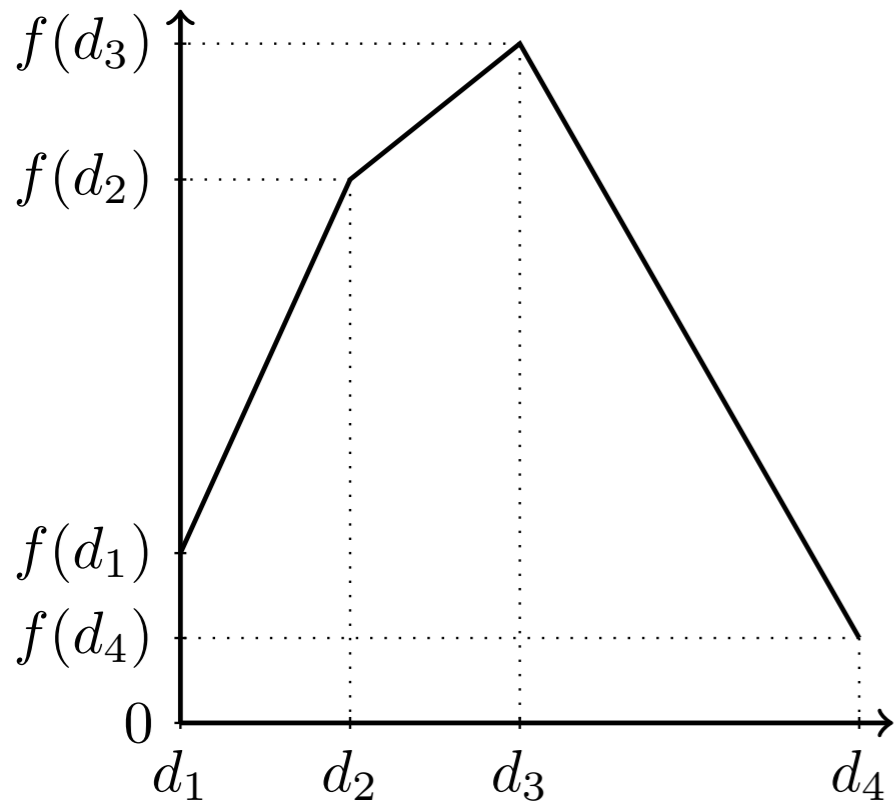
$$\lambda_1 \leq y_1,$$

$$\lambda_2 \leq y_1 + y_2$$

$$\lambda_3 \leq y_2 + y_3,$$

$$\lambda_4 \leq y_3$$

# Textbook Formulation



Formulation for  $f(x)=z$

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$$\sum_{i=1}^3 y_i = 1,$$

$$y_i \in \{0, 1\}$$

$$\lambda_1 \leq y_1,$$

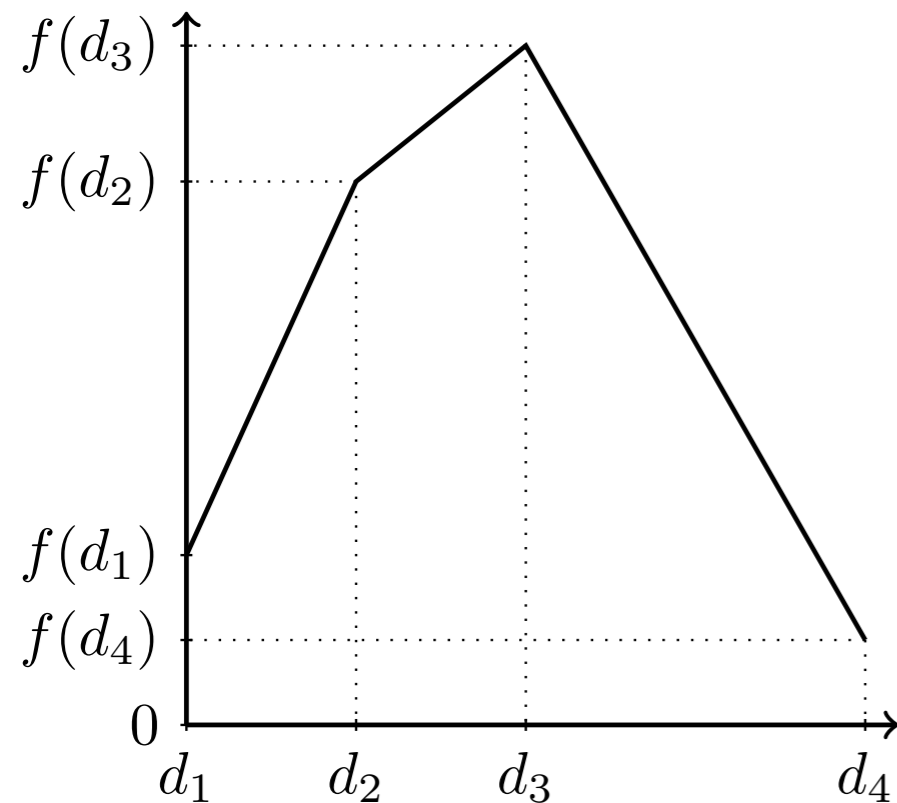
$$\lambda_2 \leq y_1 + y_2$$

$$\lambda_3 \leq y_2 + y_3,$$

$$\lambda_4 \leq y_3$$

**“Weak”**

# Better Formulation



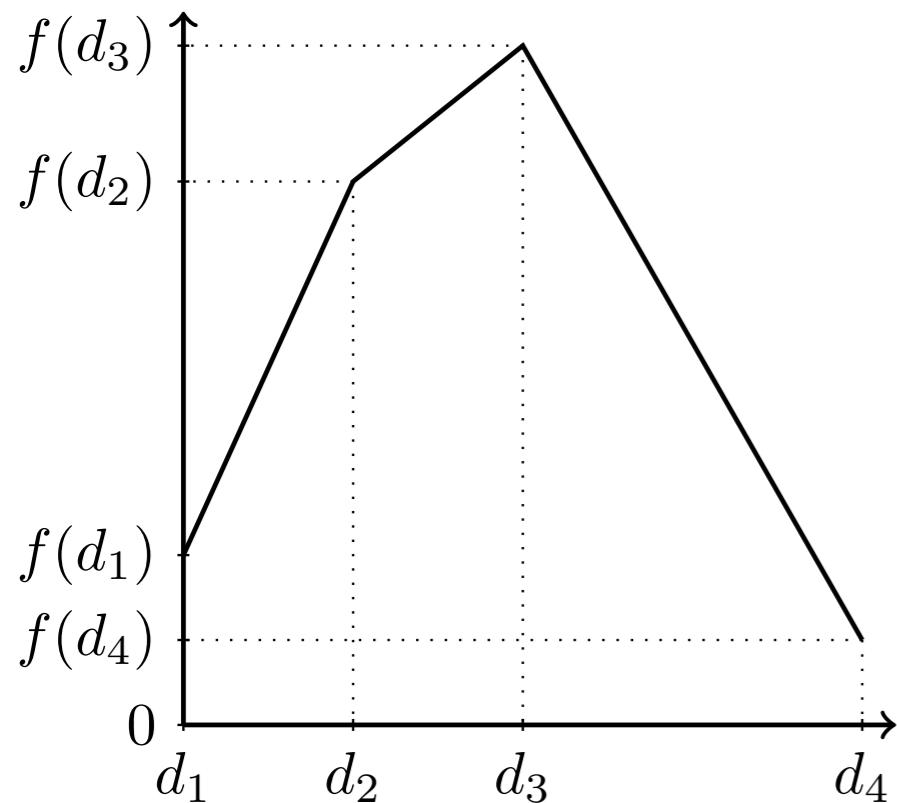
Formulation for  $f(x)=z$

$$d_0 + \sum_{i=1}^3 (d_{i+1} - d_i) \delta_i = x,$$

$$f(d_0) + \sum_{i=1}^3 (f(d_{i+1}) - f(d_i)) \delta_i = z$$

$$\delta_3 \leq y_2 \leq \delta_2 \leq y_1 \leq \delta_1$$
$$y_i \in \{0, 1\}$$

# Better Formulation



Formulation for  $f(x)=z$

$$d_0 + \sum_{i=1}^3 (d_{i+1} - d_i) \delta_i = x,$$

$$f(d_0) + \sum_{i=1}^3 (f(d_{i+1}) - f(d_i)) \delta_i = z$$

**Integral**

$$\delta_3 \leq y_2 \leq \delta_2 \leq y_1 \leq \delta_1$$

$$y_i \in \{0, 1\}$$

# Solve Times in CPLEX 11

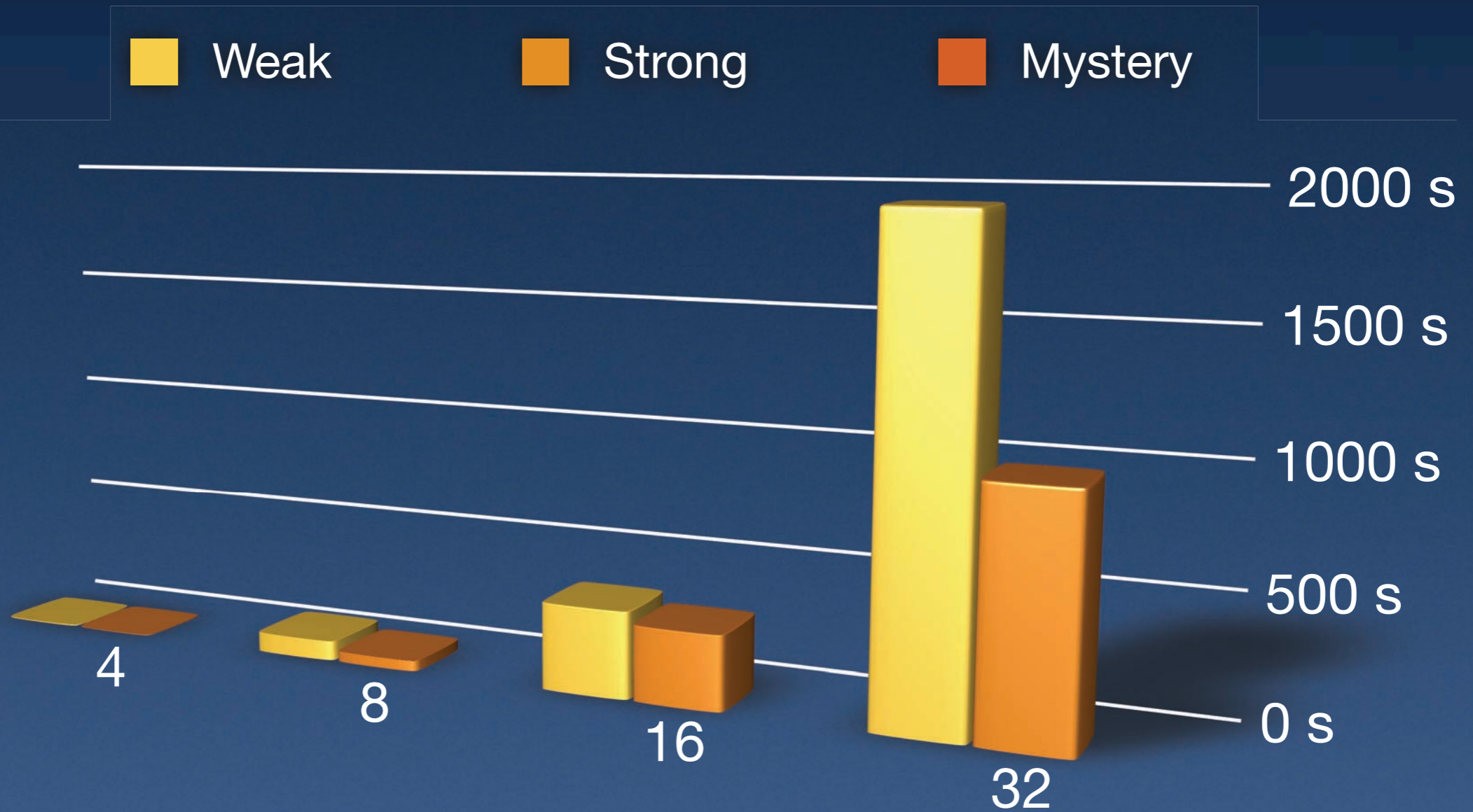
■ Weak

■ Strong

■ Mystery

- Transportation Problems in V., Ahmed and Nemhauser '10.

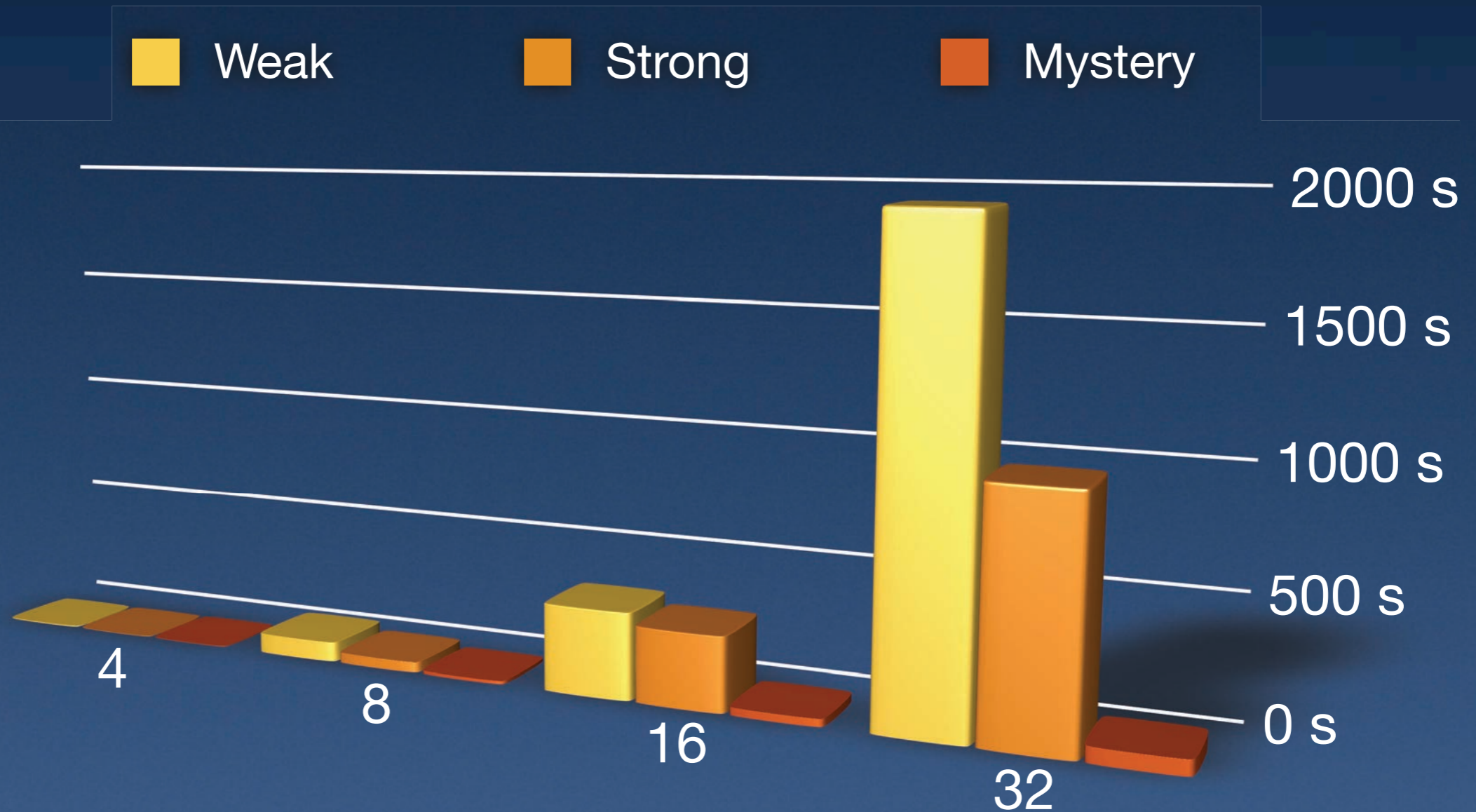
# Solve Times in CPLEX 11



- Transportation Problems in V., Ahmed and Nemhauser '10.



# Solve Times in CPLEX 11



- Transportation Problems in V., Ahmed and Nemhauser '10.

# Outline

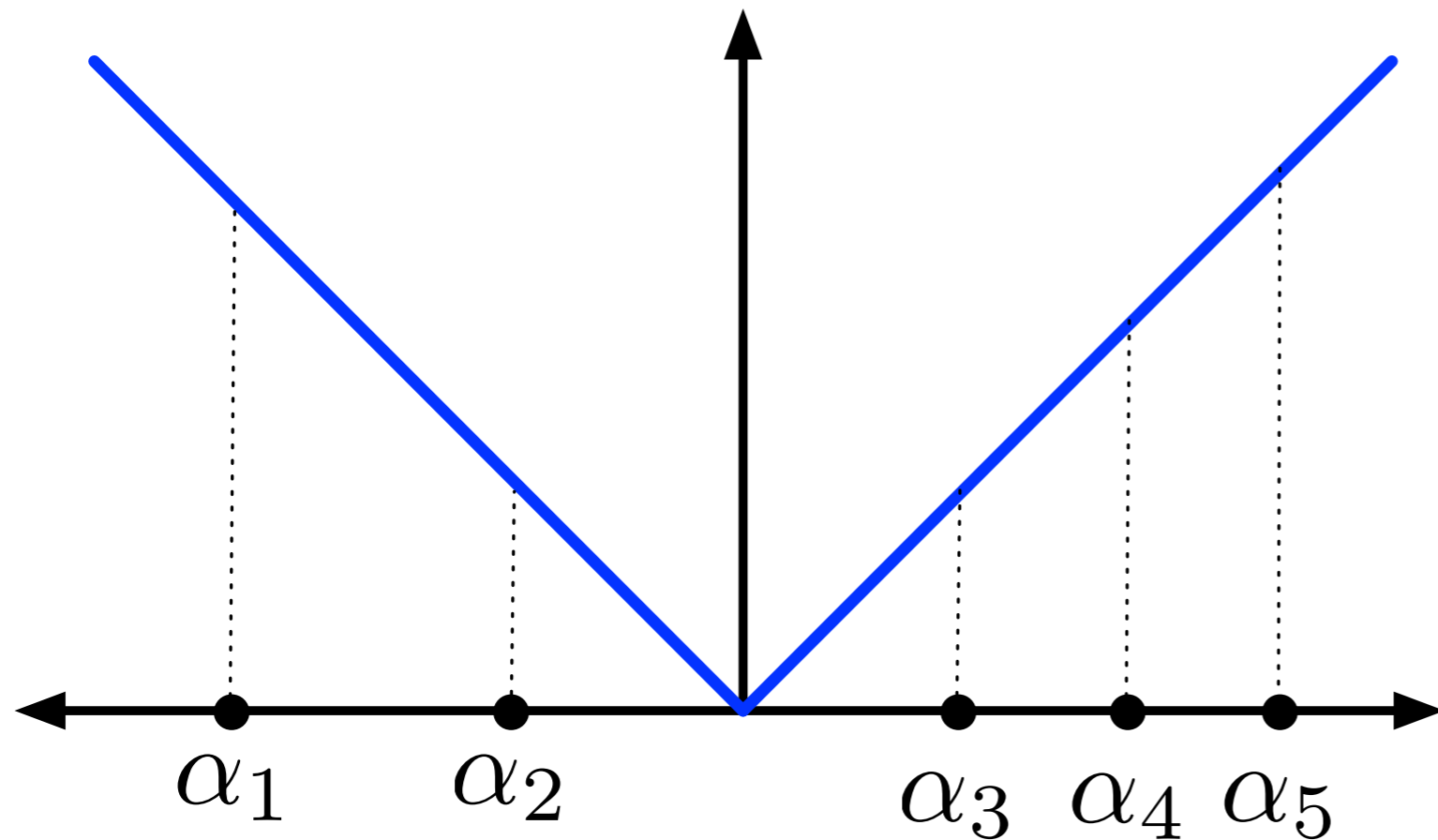
- MIP v/s constraint branching.
- “Have your cake and eat it too” formulation
  - Step 1: Encoding alternatives.
  - Step 2: Combine with strong “standard” formulation.
- Summary, Extensions and More.

# Formulating Discrete Alternatives

$$\min |x|$$

*s.t.*

$$x \in \{\alpha_i\}_{i=1}^n$$

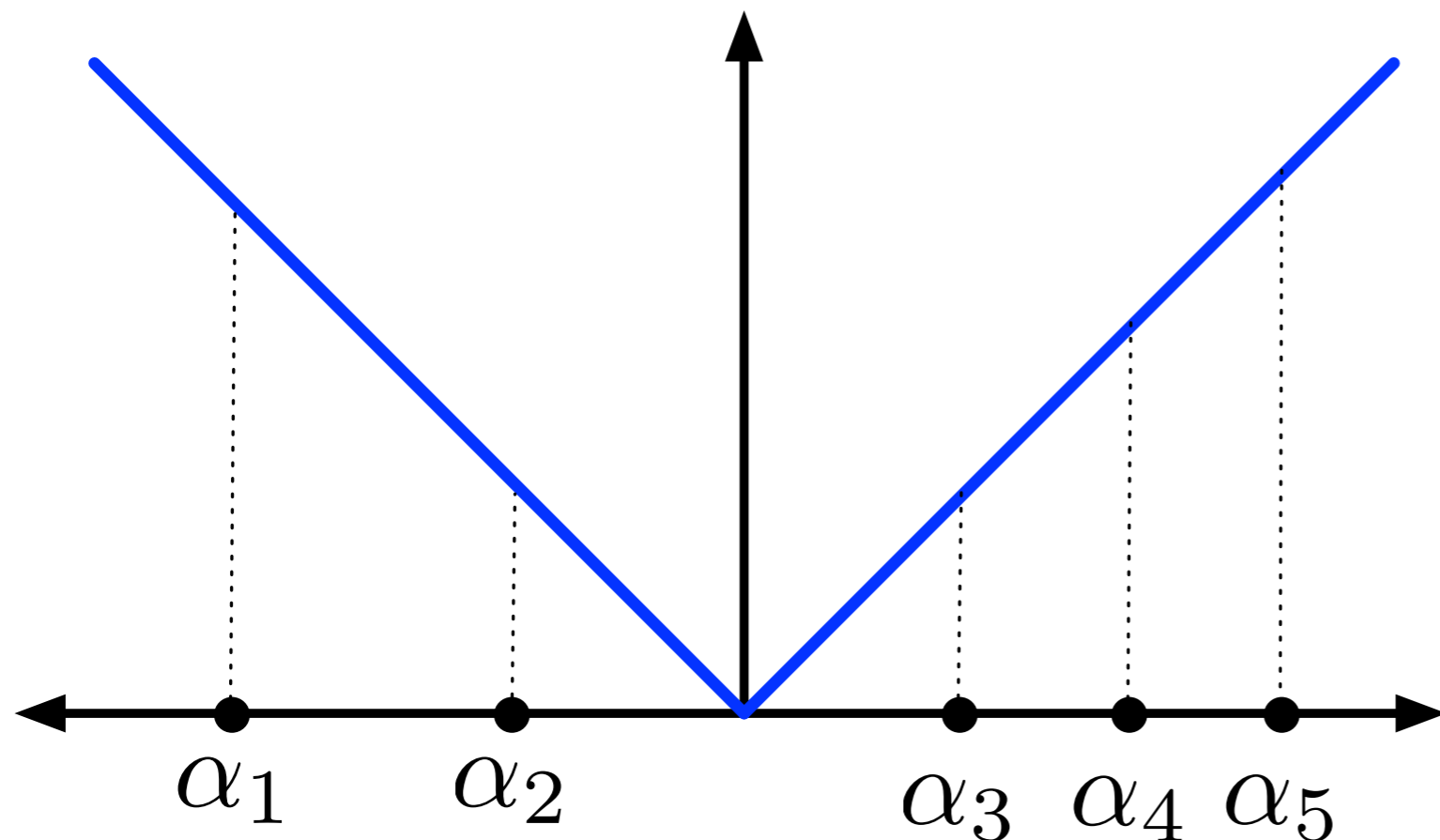


# Formulating Discrete Alternatives

$$\min |x|$$

*s.t.*

$$x \in \{\alpha_i\}_{i=1}^n$$



$$\min |x|$$

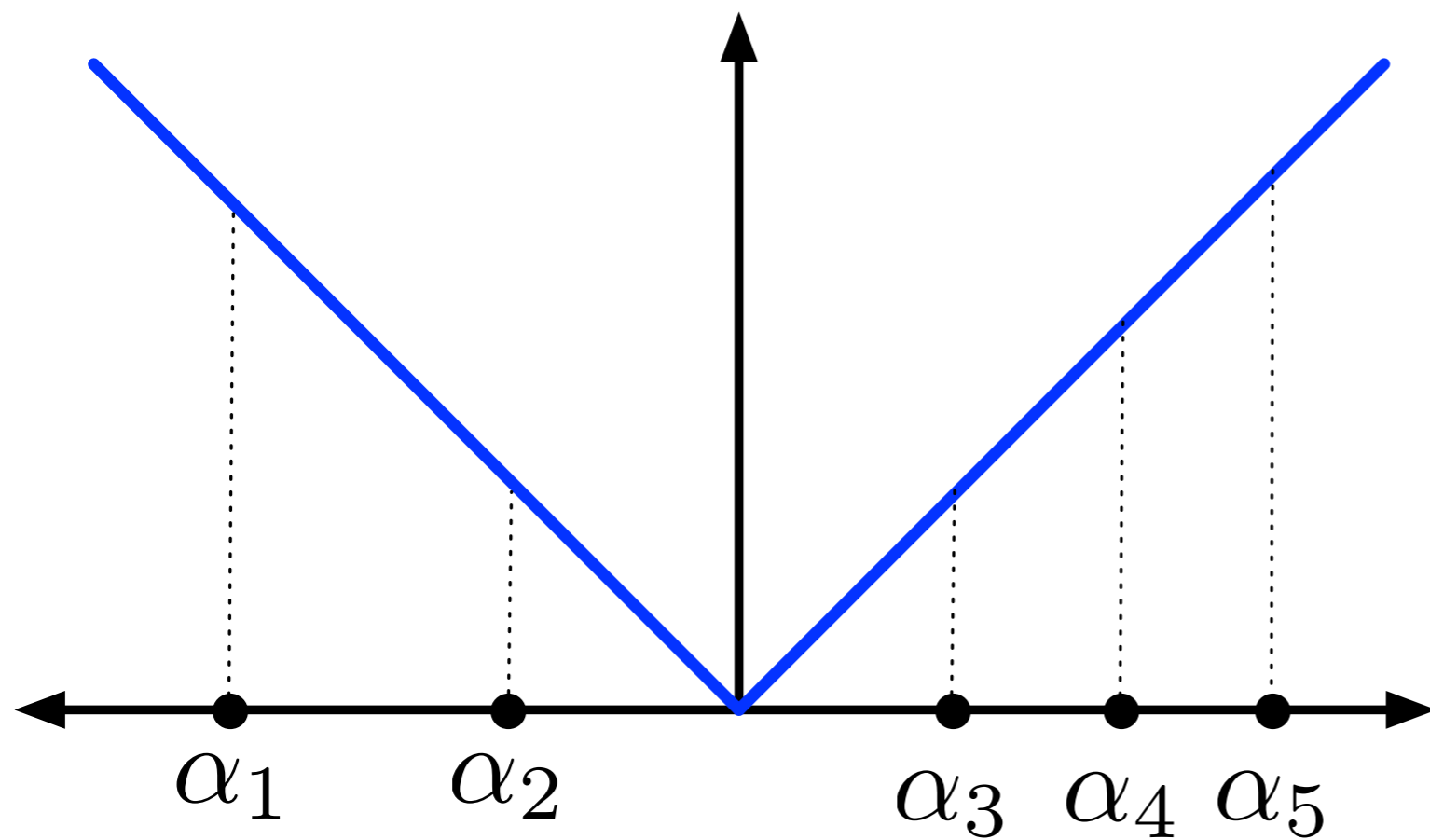
*s.t.*

$$\sum_{i=1}^n \lambda_i \alpha_i = x$$

$$\sum_{i=1}^n \lambda_i = 1$$

$$\lambda \in \{0, 1\}^n$$

# Formulating Discrete Alternatives



min

$|x|$

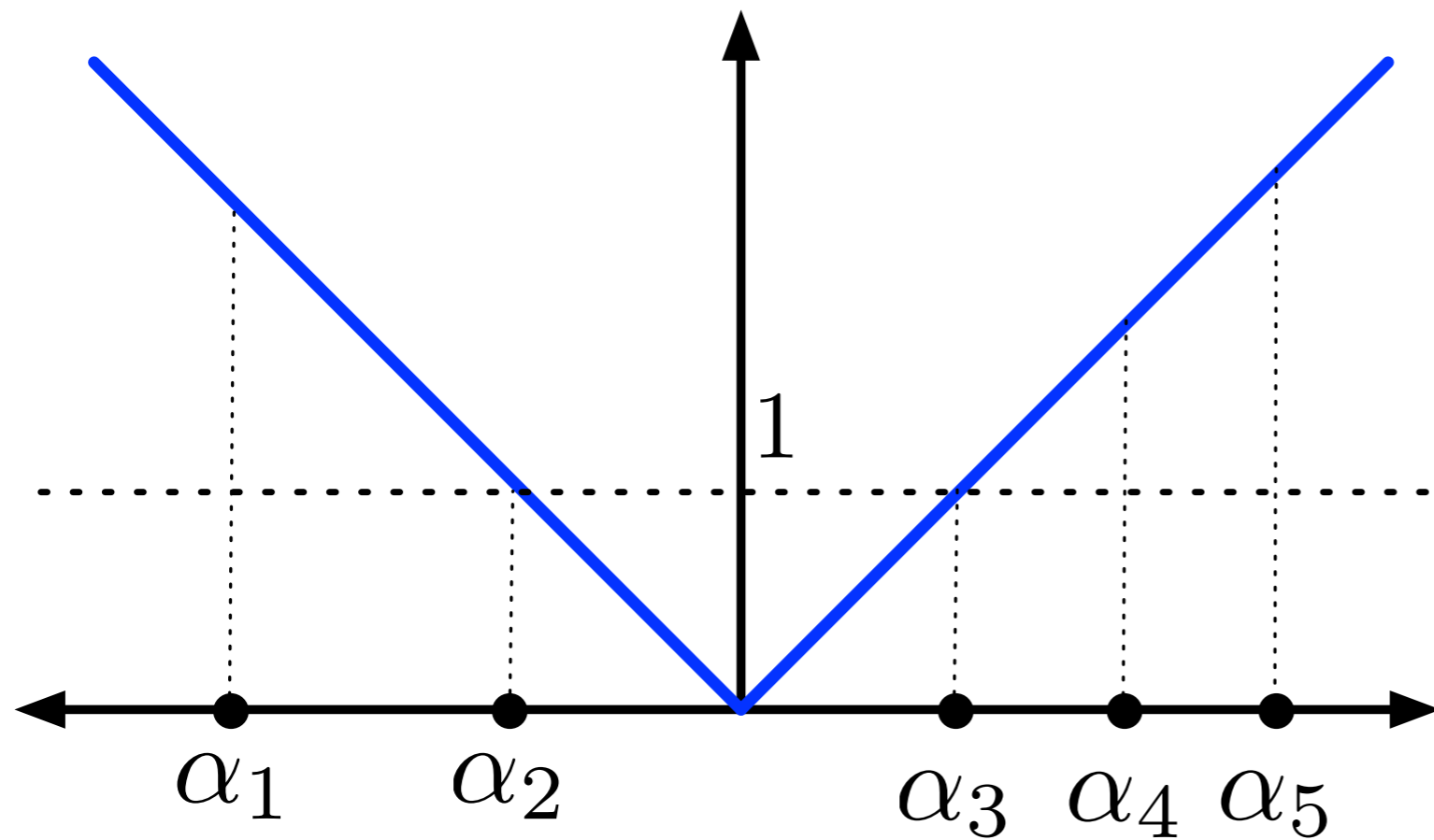
s.t.

$$\sum_{i=1}^n \lambda_i \alpha_i = x$$

$$\sum_{i=1}^n \lambda_i = 1$$

$$\lambda \in \{0, 1\}^n$$

# Formulating Discrete Alternatives



min

$|x|$

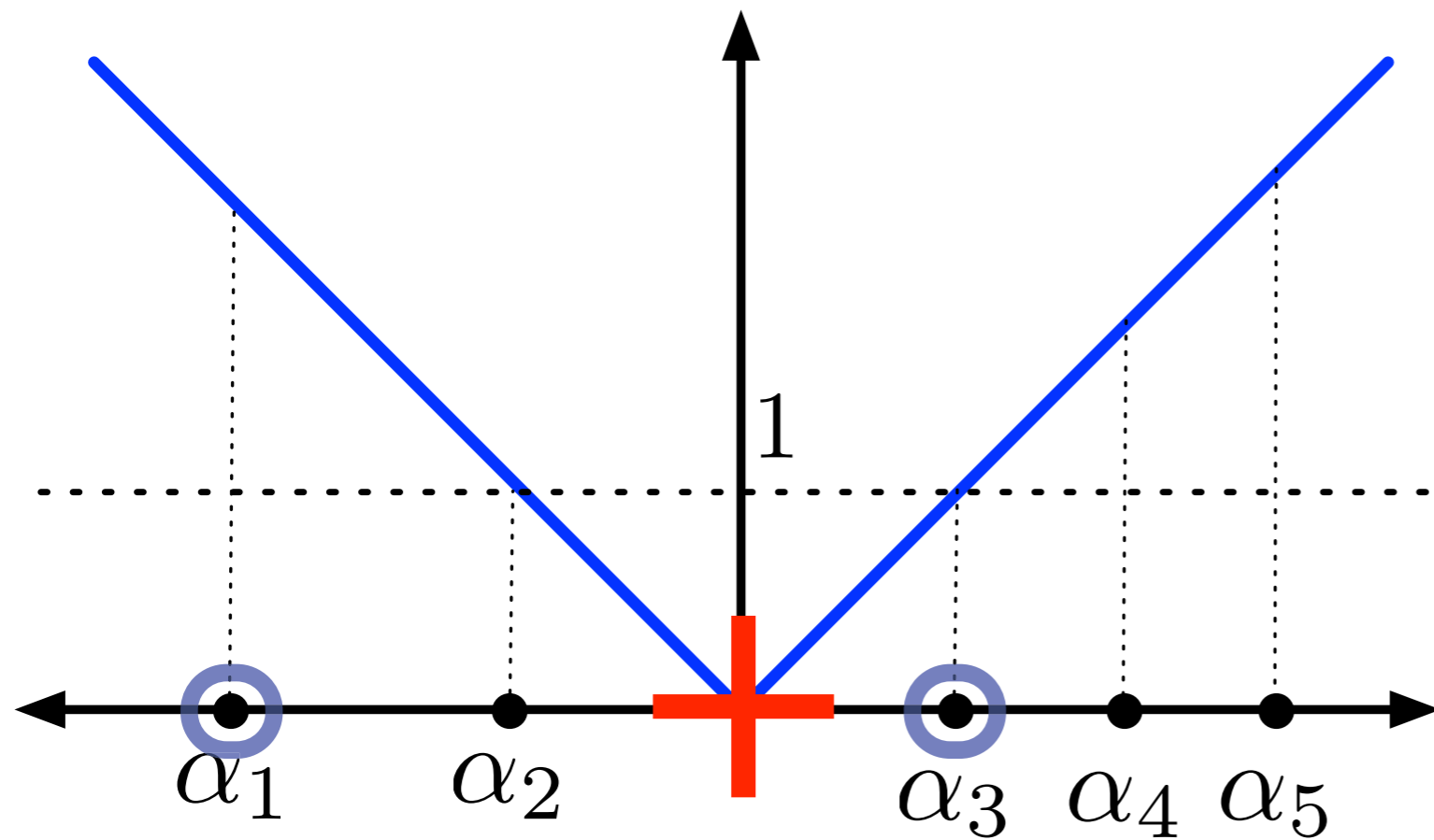
s.t.

$$\sum_{i=1}^n \lambda_i \alpha_i = x$$

$$\sum_{i=1}^n \lambda_i = 1$$

$$\lambda \in \{0, 1\}^n$$

# Formulating Discrete Alternatives



min

$|x|$

s.t.

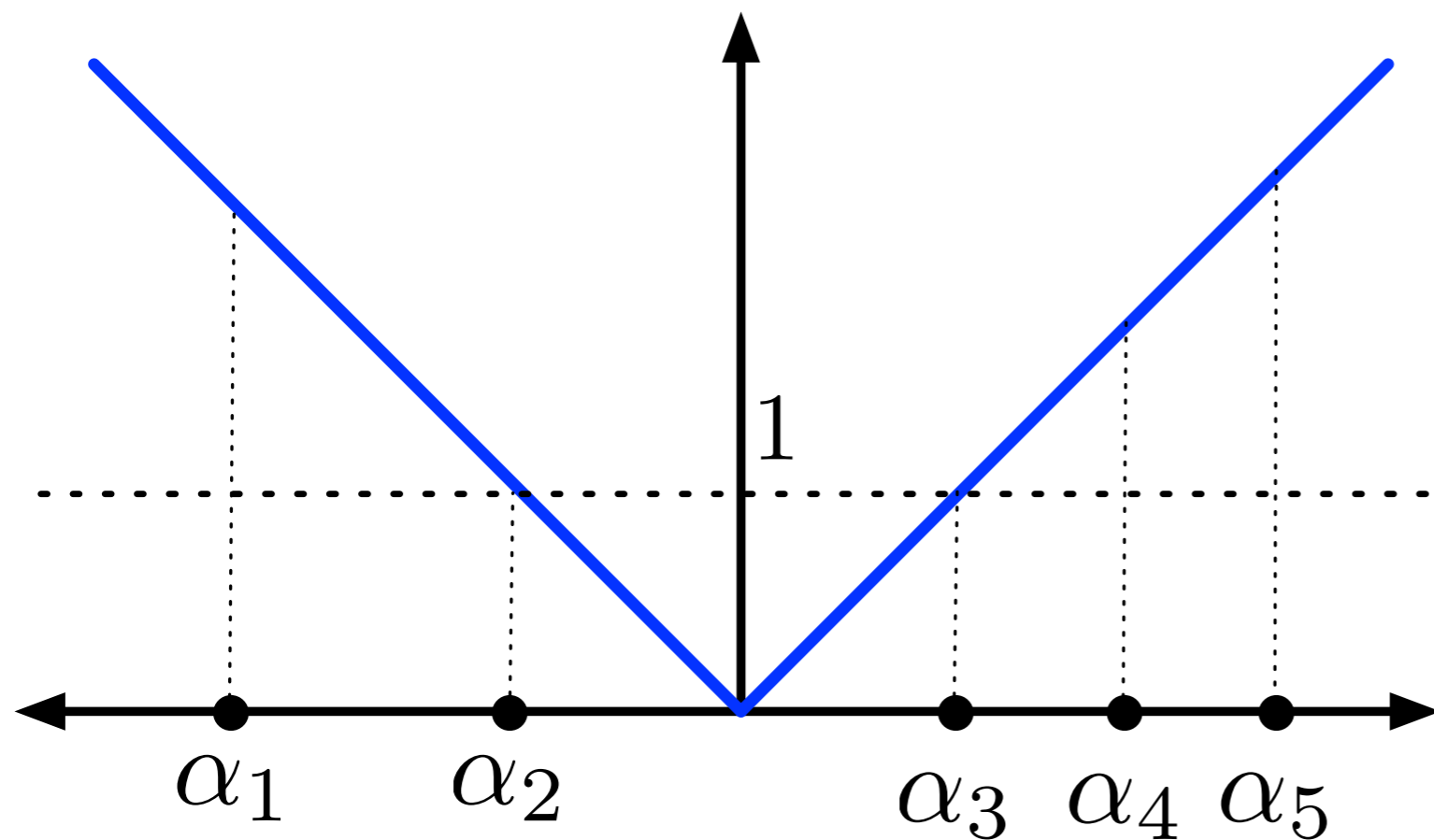
$$\sum_{i=1}^n \lambda_i \alpha_i = x$$

$$\sum_{i=1}^n \lambda_i = 1$$

$$\lambda \in \{0, 1\}^n$$

$$\text{IP}_{\text{opt}} = 1, \text{LP}_{\text{opt}} = 0$$

# Formulating Discrete Alternatives



Solve by **binary** Branch-and-Bound:

min

$|x|$

s.t.

$$\sum_{i=1}^n \lambda_i \alpha_i = x$$

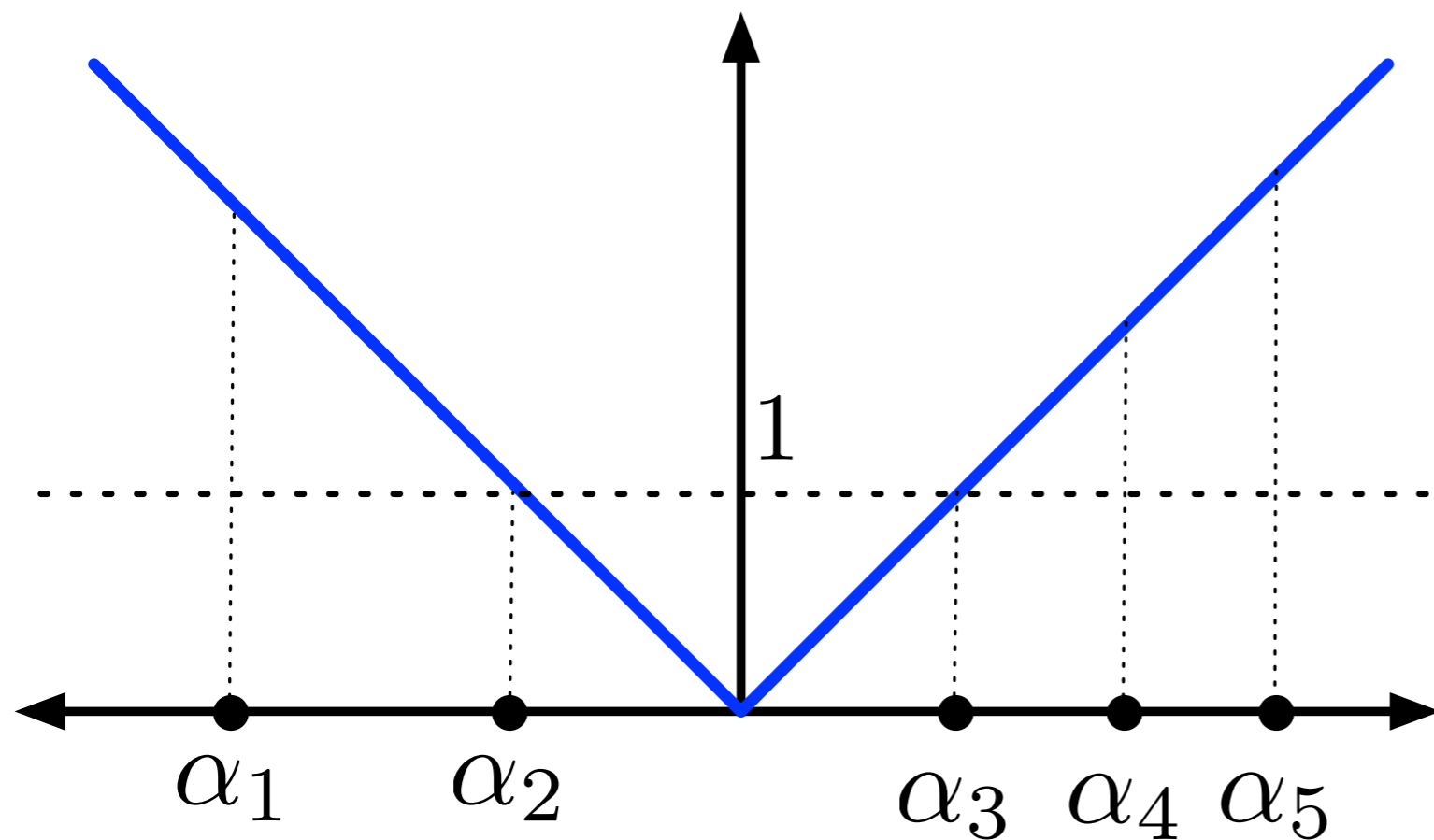
$$\sum_{i=1}^n \lambda_i = 1$$

$$\lambda \in \{0, 1\}^n$$

$$\text{IP}_{\text{opt}} = 1, \text{LP}_{\text{opt}} = 0$$



# Formulating Discrete Alternatives



min

$|x|$

s.t.

$$\sum_{i=1}^n \lambda_i \alpha_i = x$$

$$\sum_{i=1}^n \lambda_i = 1$$

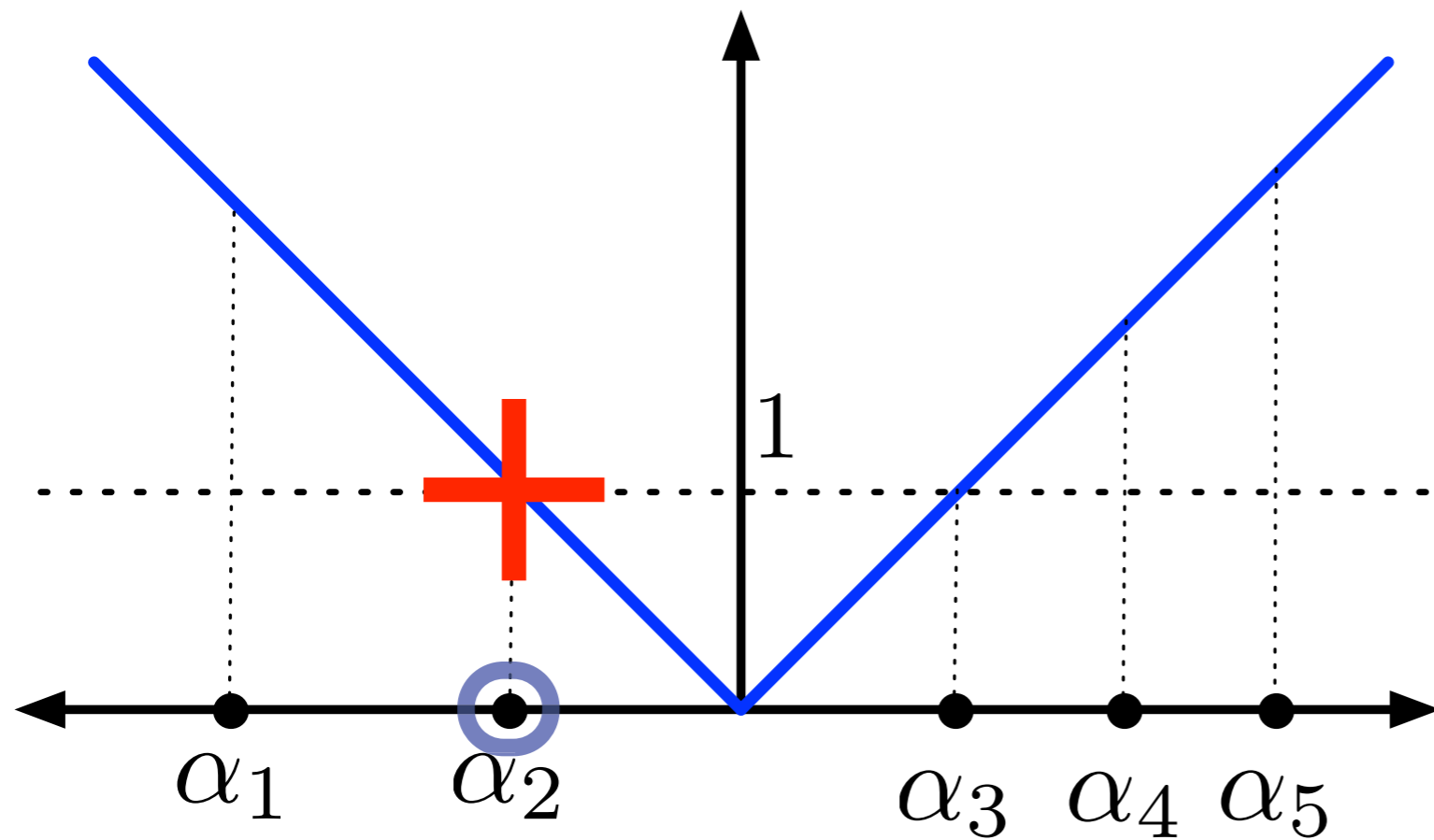
$$\lambda \in \{0, 1\}^n$$

$$\text{IP}_{\text{opt}} = 1, \text{LP}_{\text{opt}} = 0$$

Solve by **binary** Branch-and-Bound:

Branch on  $\lambda_2$

# Formulating Discrete Alternatives



min

$|x|$

s.t.

$$\sum_{i=1}^n \lambda_i \alpha_i = x$$

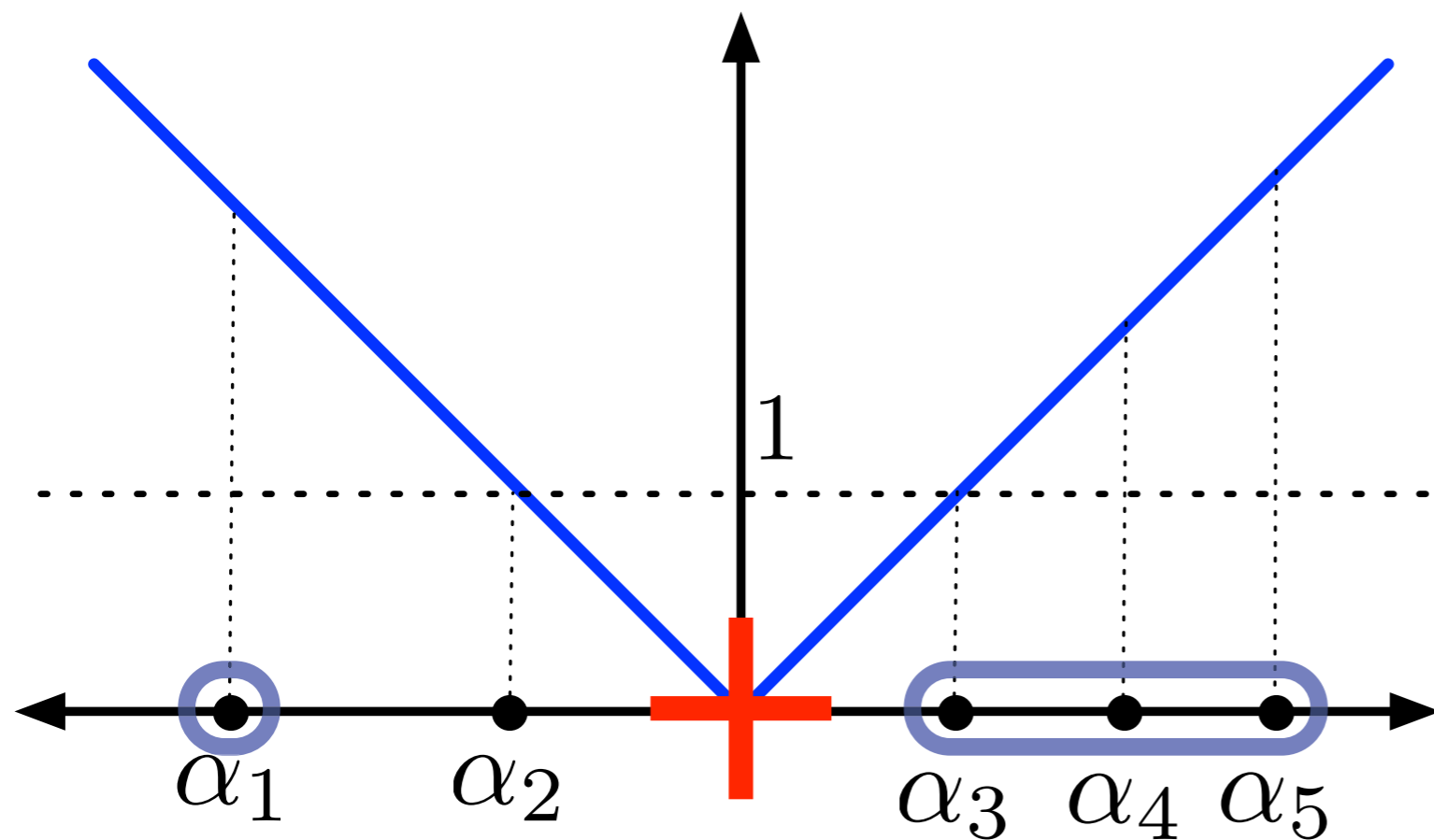
$$\sum_{i=1}^n \lambda_i = 1$$

$$\lambda \in \{0, 1\}^n$$

Solve by **binary** Branch-and-Bound:  $IP_{\text{opt}} = 1, LP_{\text{opt}} = 0$

Branch on  $\lambda_2$   $\rightarrow$   $\bullet \lambda_2 = 1 \rightarrow$  Feasible with  $|x| = 1$

# Formulating Discrete Alternatives



min

$|x|$

s.t.

$$\sum_{i=1}^n \lambda_i \alpha_i = x$$

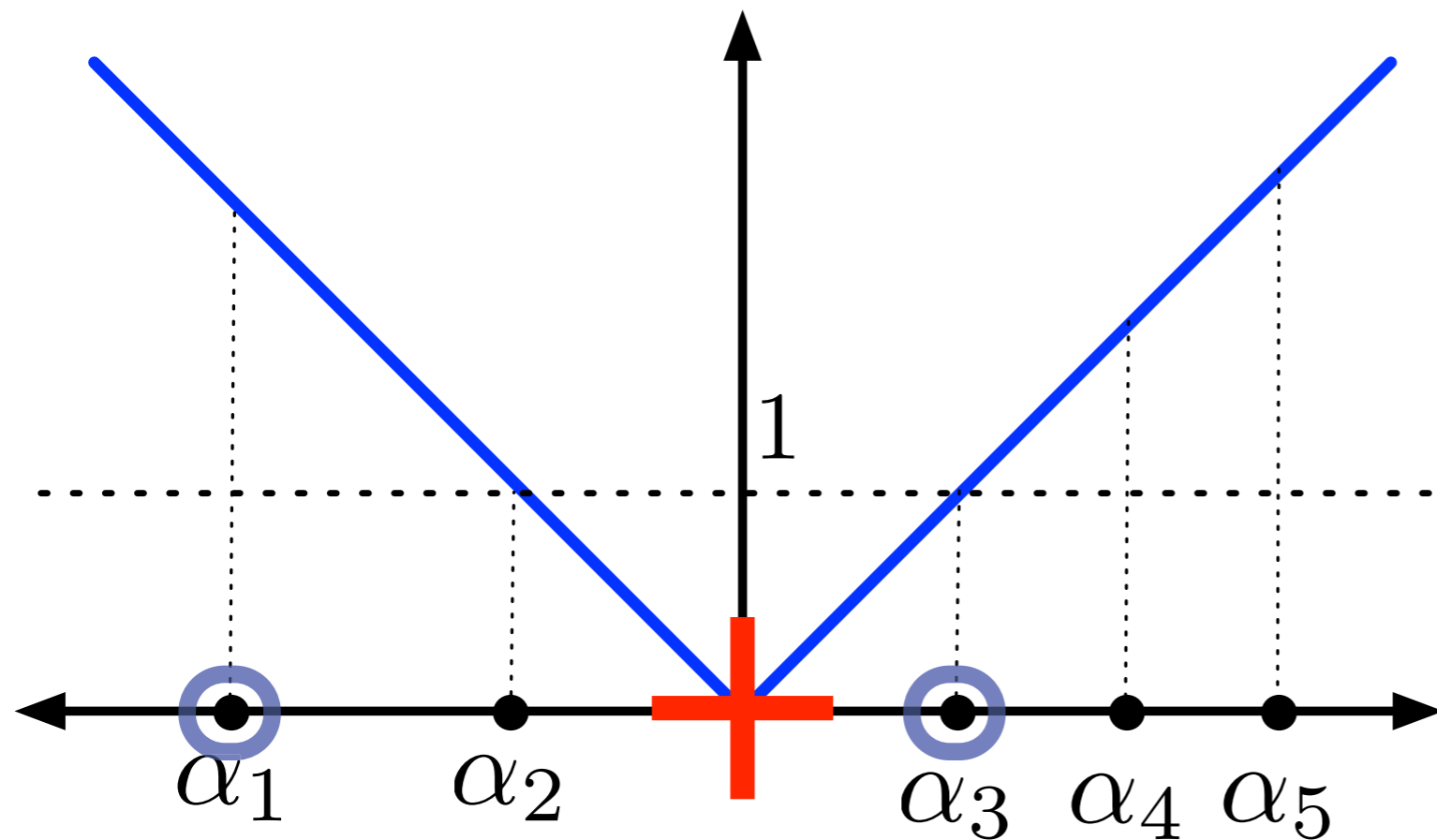
$$\sum_{i=1}^n \lambda_i = 1$$

$$\lambda \in \{0, 1\}^n$$

Solve by **binary** Branch-and-Bound:  $IP_{\text{opt}} = 1, LP_{\text{opt}} = 0$

- Branch on  $\lambda_2$
- $\lambda_2 = 1 \rightarrow$  Feasible with  $|x| = 1$
  - $\lambda_2 = 0 \rightarrow$  Best Bound = 0

# Formulating Discrete Alternatives



min

$|x|$

s.t.

$$\sum_{i=1}^n \lambda_i \alpha_i = x$$

$$\sum_{i=1}^n \lambda_i = 1$$

$$\lambda \in \{0, 1\}^n$$

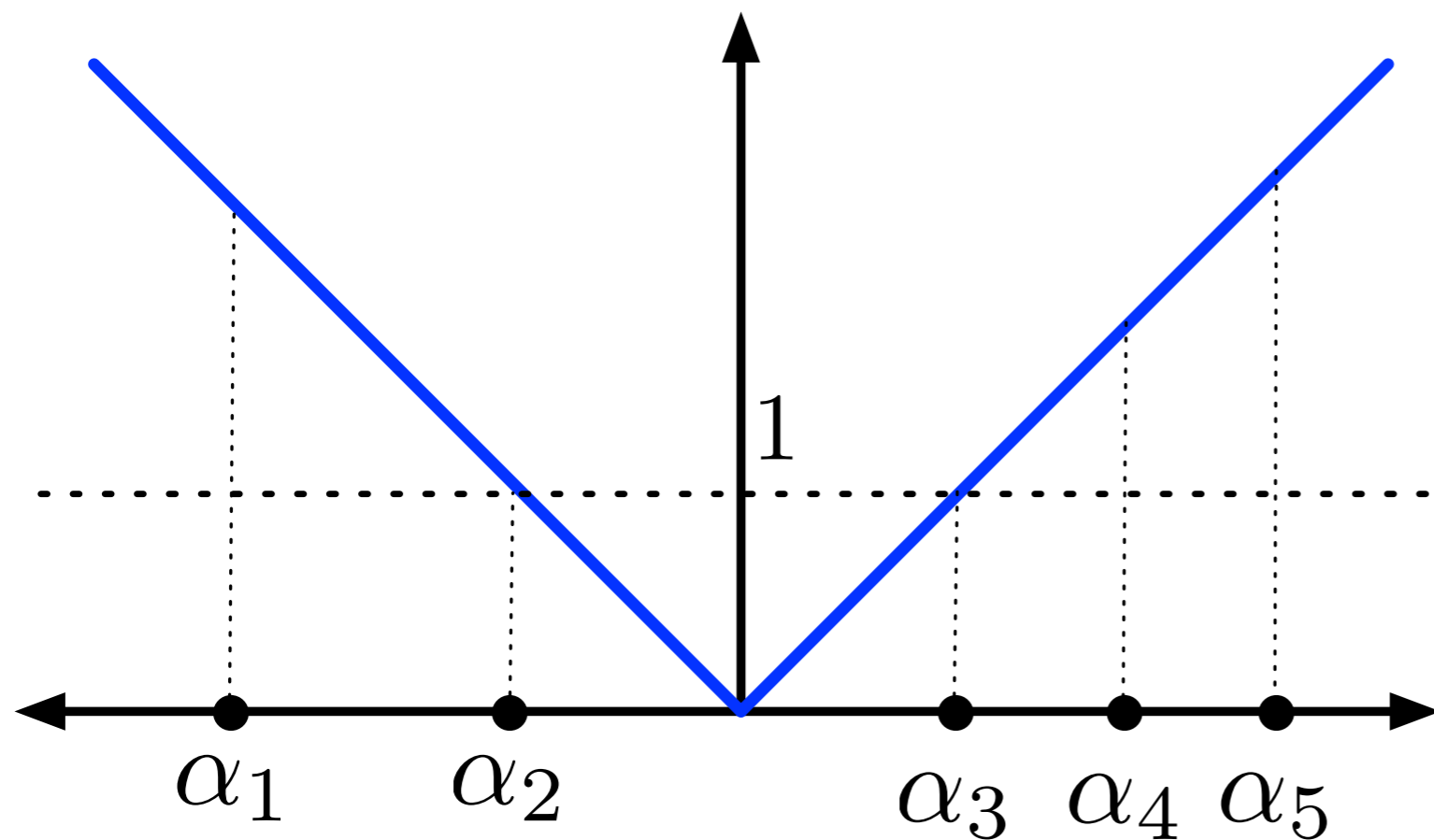
Solve by **binary** Branch-and-Bound:  $IP_{\text{opt}} = 1, LP_{\text{opt}} = 0$

Branch on  $\lambda_2$  →

- $\lambda_2 = 1 \rightarrow$  Feasible with  $|x| = 1$
- $\lambda_2 = 0 \rightarrow$  Best Bound = 0

Branch on  $\lambda_2, \lambda_4, \lambda_5 \rightarrow$  Best Bound = 0

# Formulating Discrete Alternatives



min

$|x|$

s.t.

$$\sum_{i=1}^n \lambda_i \alpha_i = x$$

$$\sum_{i=1}^n \lambda_i = 1$$

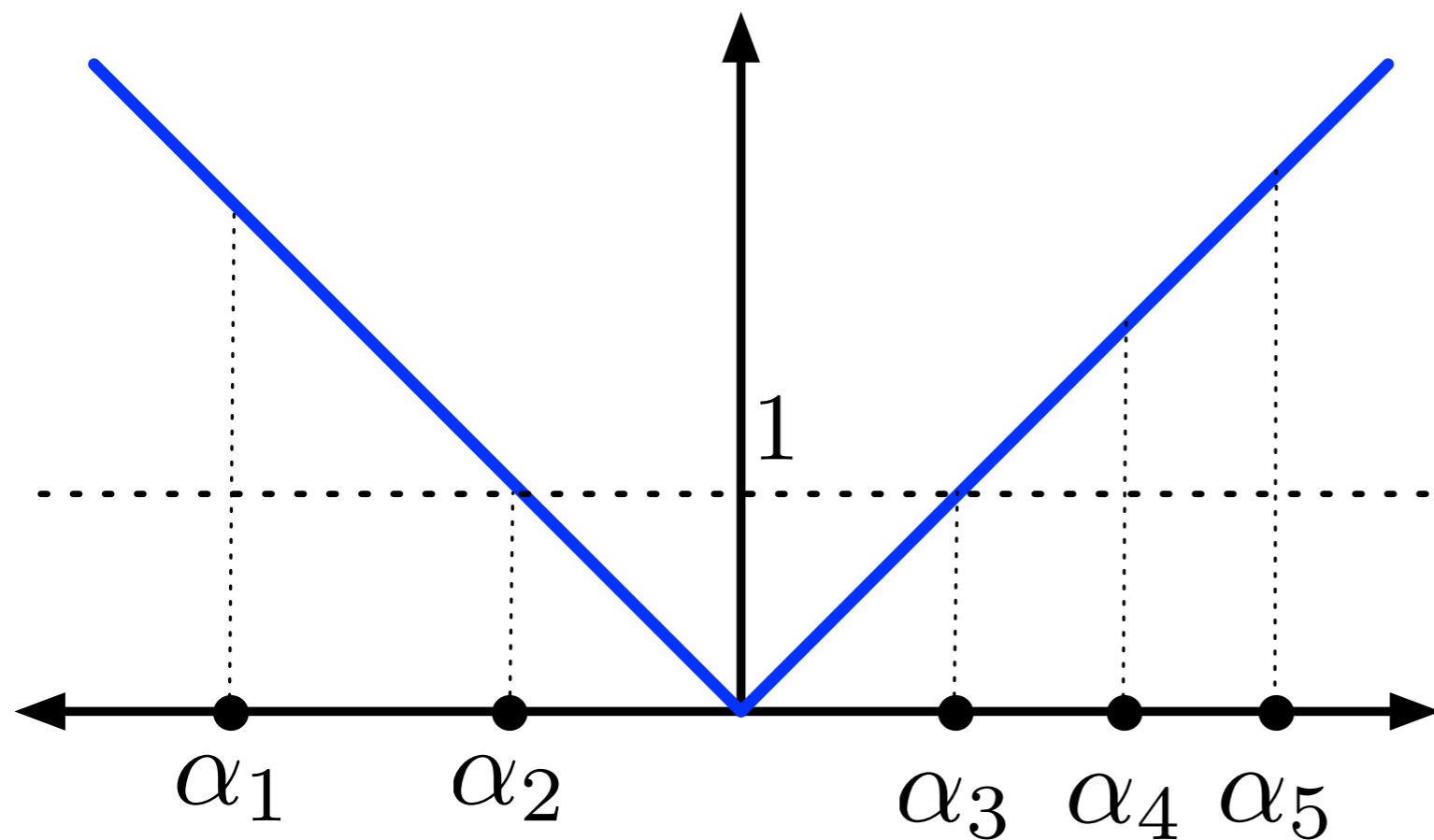
$$\lambda \in \{0, 1\}^n$$

Solve by **binary** Branch-and-Bound:

$$\text{IP}_{\text{opt}} = 1, \text{LP}_{\text{opt}} = 0$$

Worst case:  $n/2$  branches to solve

# Formulating Discrete Alternatives



Solve by **constraint** B-and-B:

min

$|x|$

s.t.

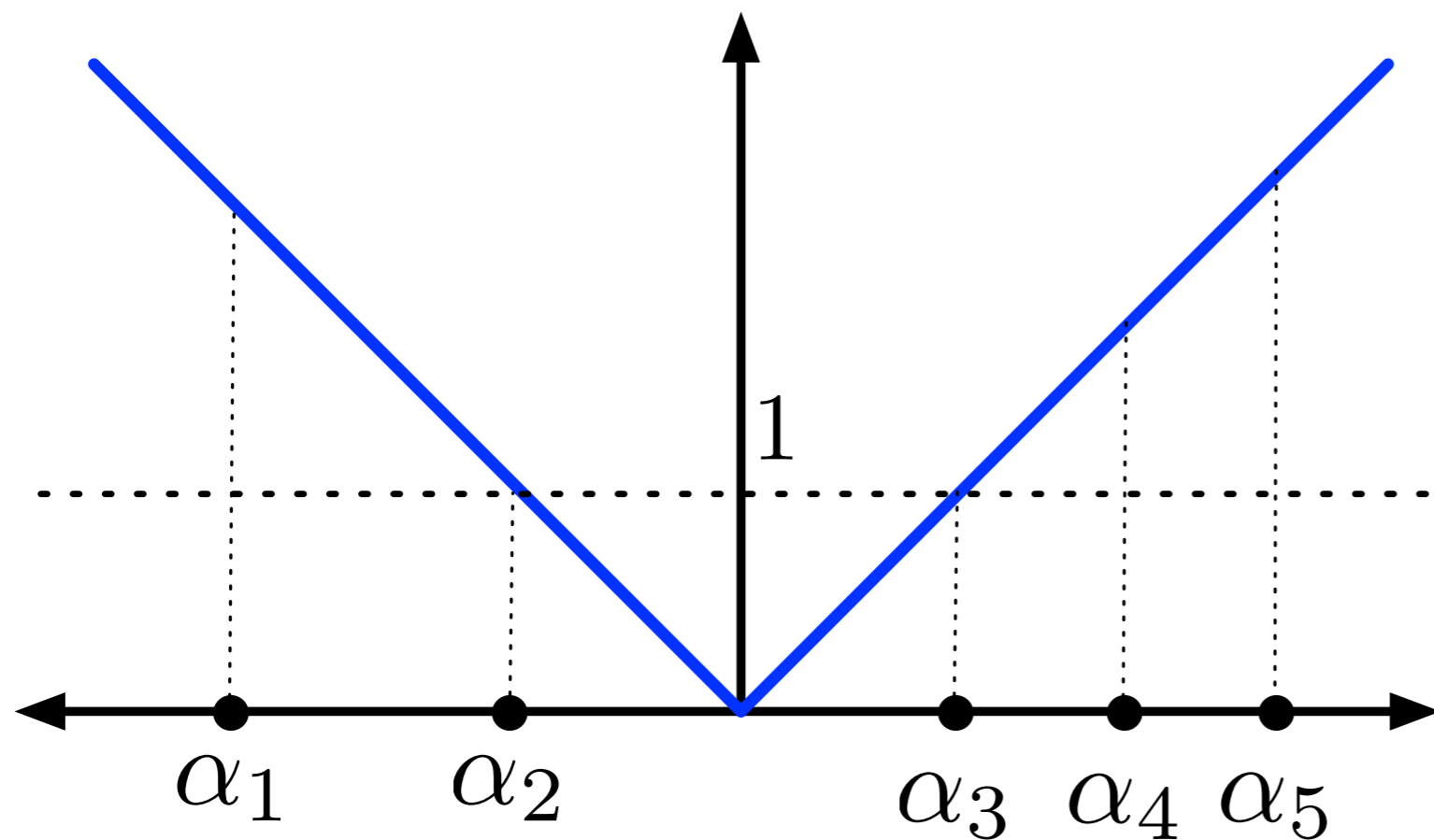
$$\sum_{i=1}^n \lambda_i \alpha_i = x$$

$$\sum_{i=1}^n \lambda_i = 1$$

$$\lambda \in \{0, 1\}^n$$

$$\text{IP}_{\text{opt}} = 1, \text{LP}_{\text{opt}} = 0$$

# Formulating Discrete Alternatives



Solve by **constraint** B-and-B:

Branch on  $\lambda_1 + \lambda_2$

min

$|x|$

s.t.

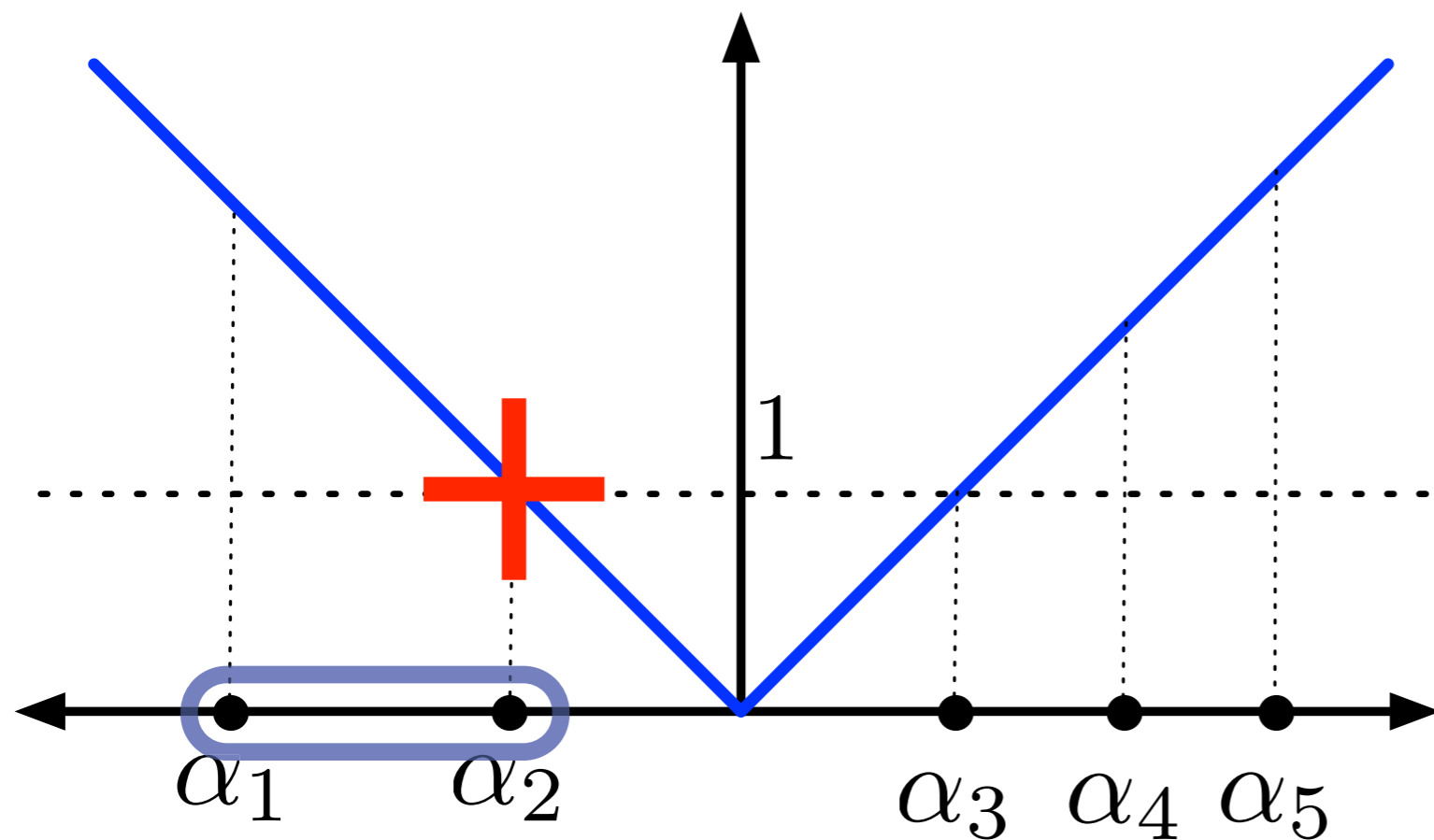
$$\sum_{i=1}^n \lambda_i \alpha_i = x$$

$$\sum_{i=1}^n \lambda_i = 1$$

$$\lambda \in \{0, 1\}^n$$

$$\text{IP}_{\text{opt}} = 1, \text{LP}_{\text{opt}} = 0$$

# Formulating Discrete Alternatives



min

$|x|$

s.t.

$$\sum_{i=1}^n \lambda_i \alpha_i = x$$

$$\sum_{i=1}^n \lambda_i = 1$$

$$\lambda \in \{0, 1\}^n$$

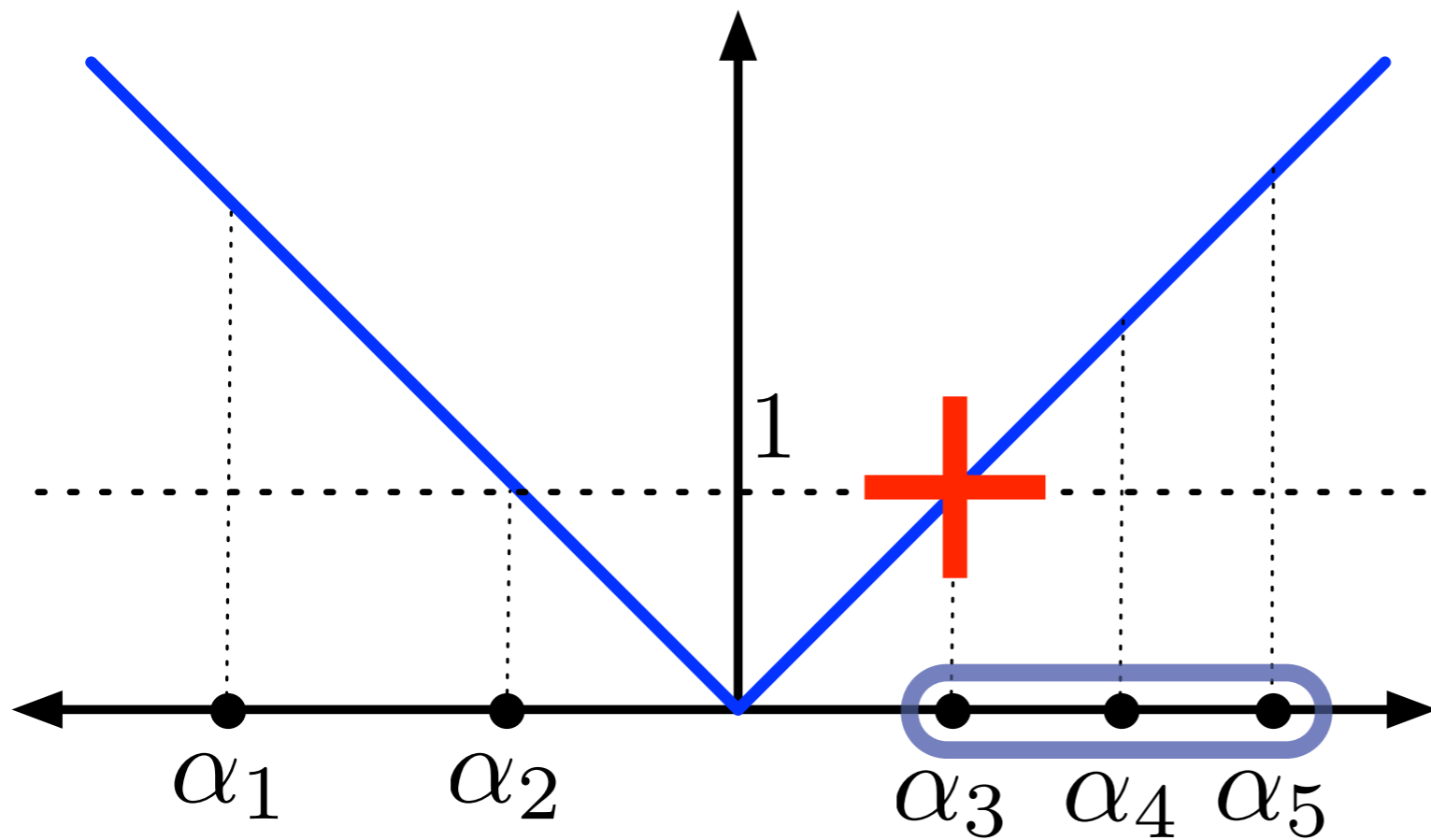
$$\text{IP}_{\text{opt}} = 1, \text{LP}_{\text{opt}} = 0$$

Solve by **constraint** B-and-B:

Branch on  $\lambda_1 + \lambda_2$  → •  $\lambda_1 + \lambda_2 = 1 \rightarrow$  Feasible with  $|x| = 1$



# Formulating Discrete Alternatives



min

$$|x|$$

s.t.

$$\sum_{i=1}^n \lambda_i \alpha_i = x$$

$$\sum_{i=1}^n \lambda_i = 1$$

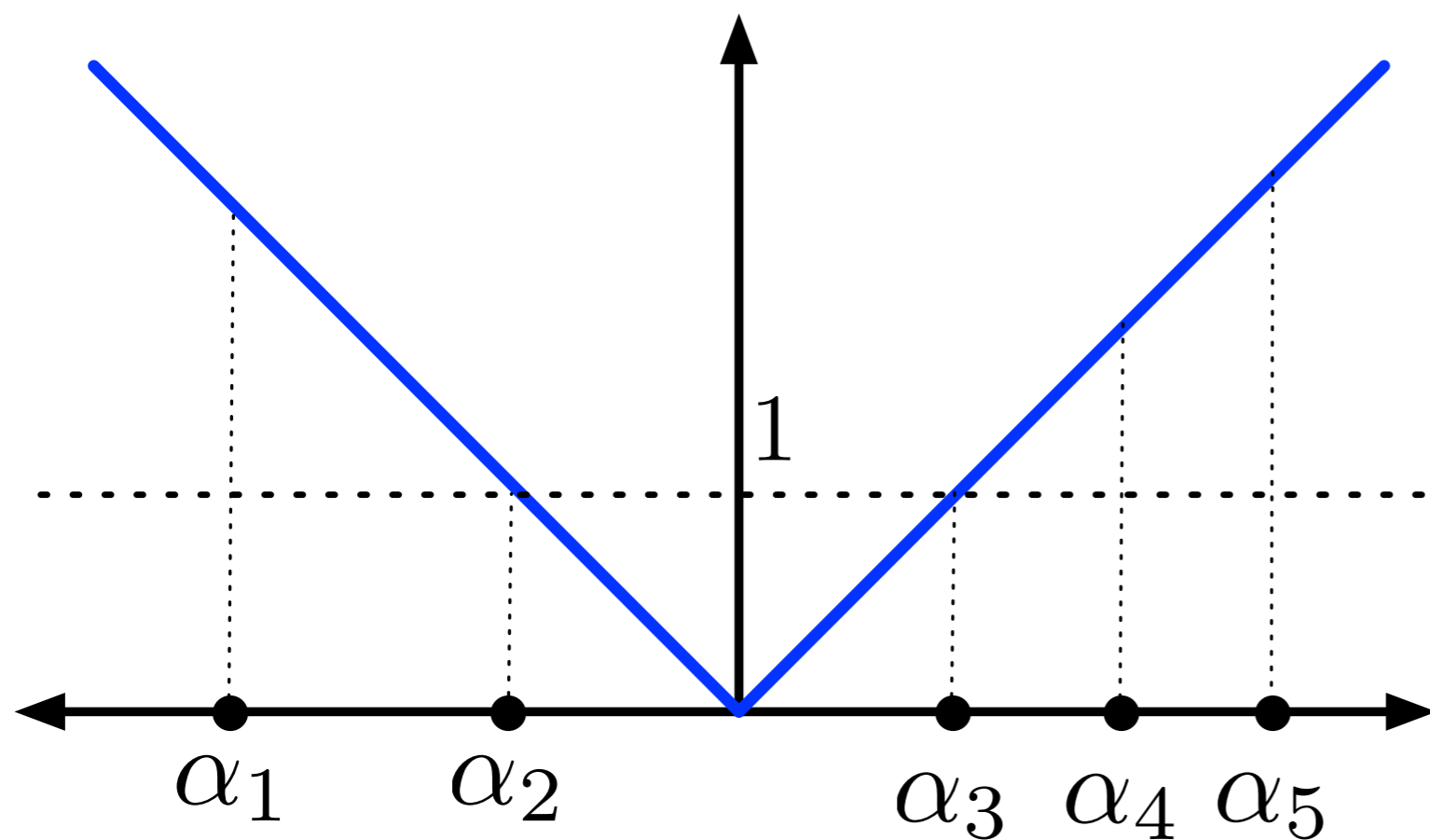
$$\lambda \in \{0, 1\}^n$$

$$IP_{\text{opt}} = 1, LP_{\text{opt}} = 0$$

Solve by **constraint** B-and-B:

- Branch on  $\lambda_1 + \lambda_2$
- $\lambda_1 + \lambda_2 = 1 \rightarrow$  Feasible with  $|x| = 1$
  - $\lambda_1 + \lambda_2 = 0 \rightarrow$  Feasible with  $|x| = 1$

# Formulating Discrete Alternatives



min

$|x|$

s.t.

$$\sum_{i=1}^n \lambda_i \alpha_i = x$$

$$\sum_{i=1}^n \lambda_i = 1$$

$$\lambda \in \{0, 1\}^n$$

$$\text{IP}_{\text{opt}} = 1, \text{LP}_{\text{opt}} = 0$$

Solve by **constraint** B-and-B:

Branch on  $\lambda_1 + \lambda_2$

- $\lambda_1 + \lambda_2 = 1 \rightarrow$  Feasible with  $|x| = 1$
- $\lambda_1 + \lambda_2 = 0 \rightarrow$  Feasible with  $|x| = 1$

Never more than one branch (2 nodes).

# Constraint Branching is the Solution?

- Ryan and Foster, 1981.
- Discrete Alternatives: SOS1 branching of Beale and Tomlin 1970. Also SOS2 (B. and T, 70) and piecewise linear functions (Tomlin 1981).

● SOS1:  $\sum_{i=1}^t \lambda_i = 1$       or       $\sum_{i=1}^t \lambda_i = 0$

$\Updownarrow$

$\lambda_i = 0 \quad \forall i > t$       or       $\lambda_i = 0 \quad \forall i \leq t$

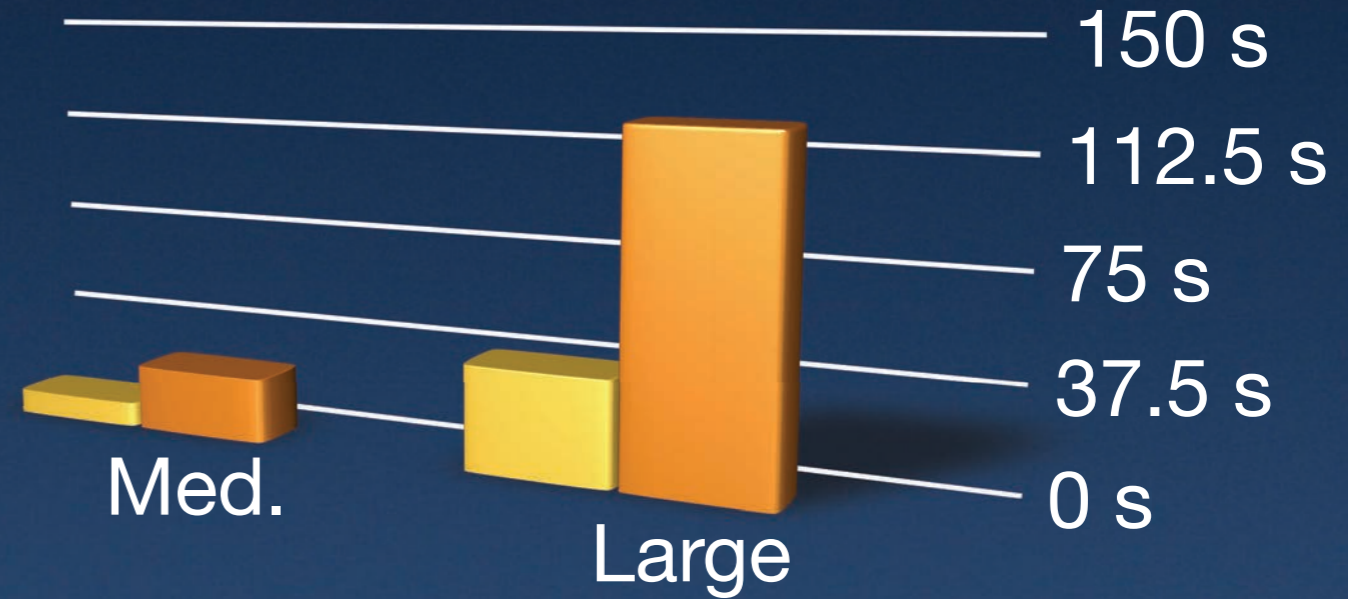
- Problem: Need to re-implement advanced branching selection (e.g. pseudocost).

# Binary v/s Specialized Branching

- Weak Integer
- SOS2 Branching
- Mystery Integer

# Binary v/s Specialized Branching

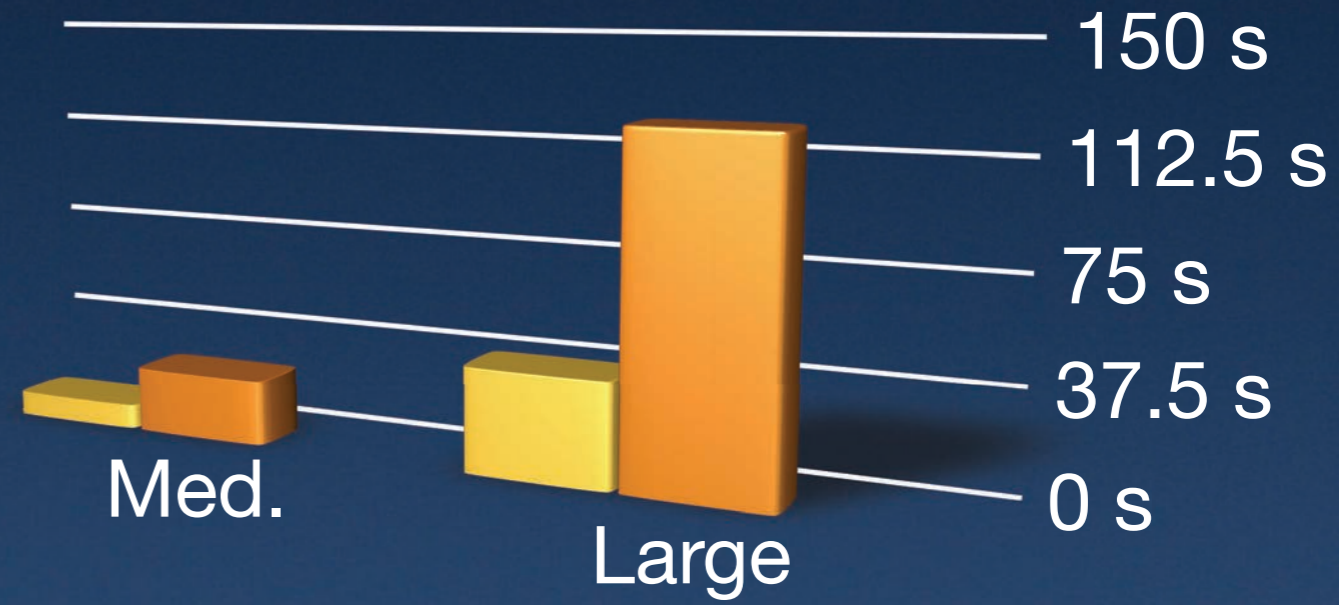
- CPLEX 9: Basic SOS2 branching implementation (graph from Nemhauser, Keha and V. '08)



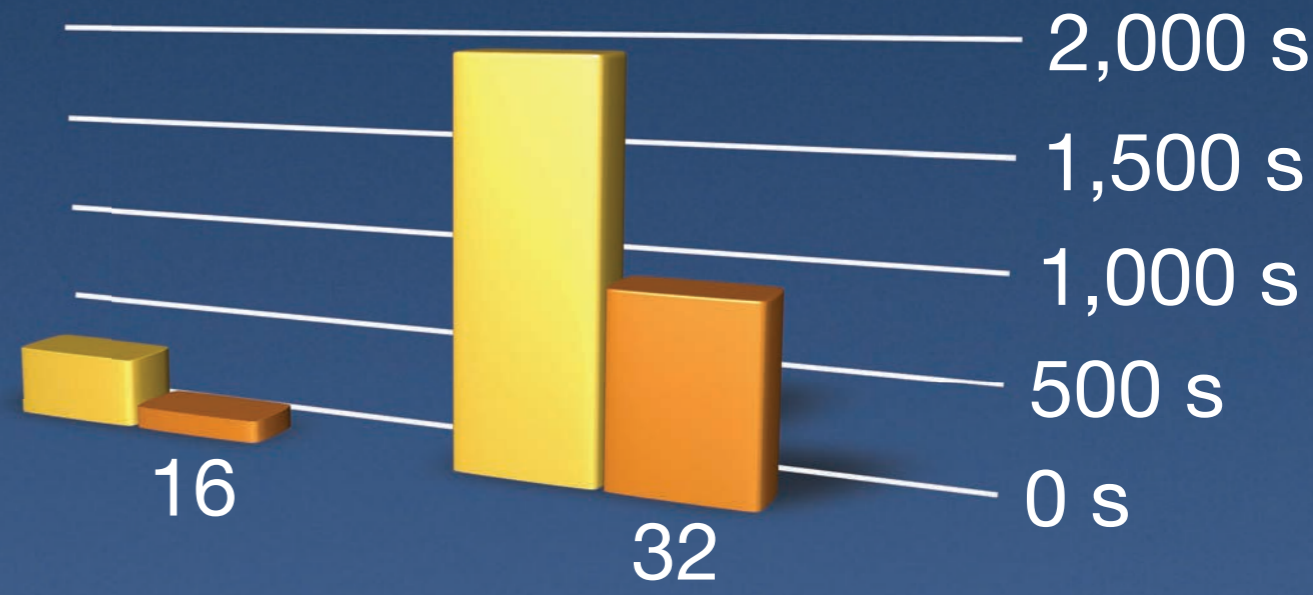
■ Weak Integer   ■ SOS2 Branching   ■ Mystery Integer

# Binary v/s Specialized Branching

- CPLEX 9: Basic SOS2 branching implementation  
(graph from Nemhauser, Keha and V. '08)



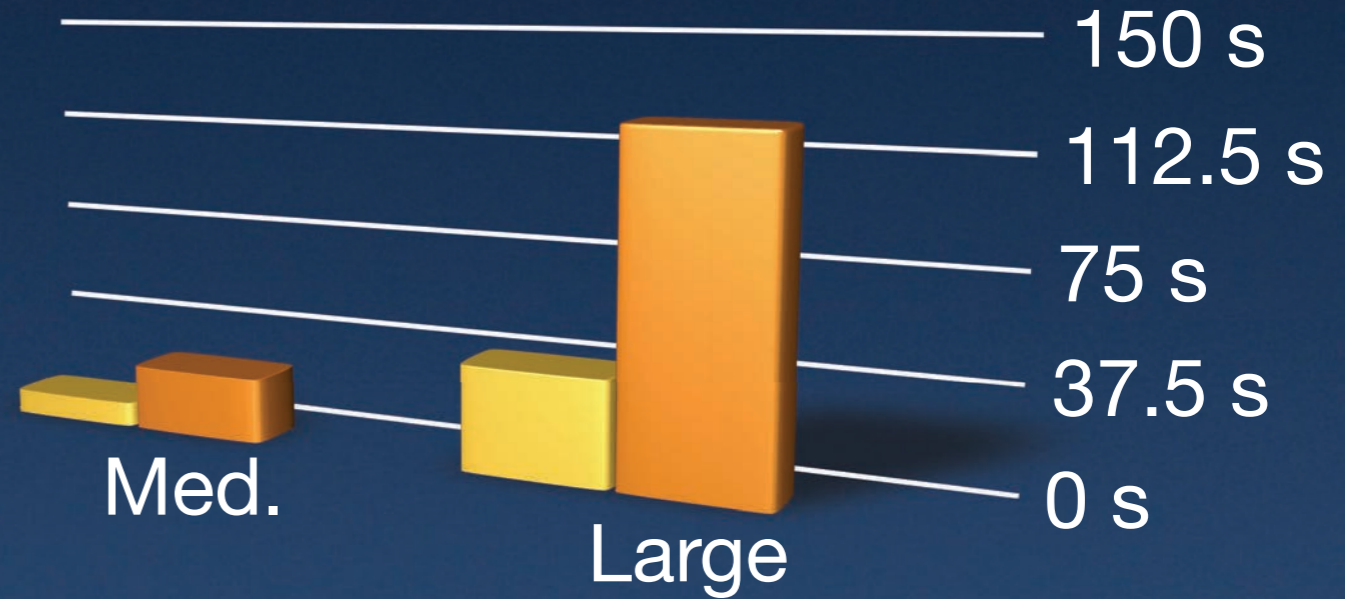
- CPLEX 11: Improved SOS2 branching implementation  
(graph from Nemhauser, Ahmed and V. '10)



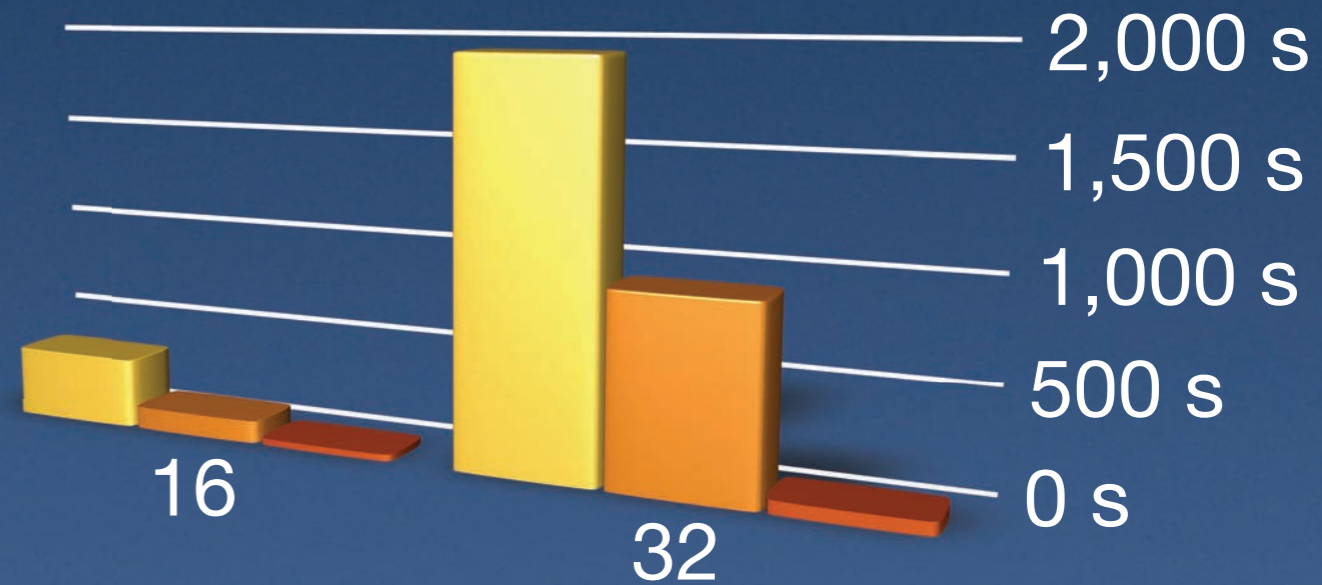
■ Weak Integer   
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# Binary v/s Specialized Branching

- CPLEX 9: Basic SOS2 branching implementation (graph from Nemhauser, Keha and V. '08)



- CPLEX 11: Improved SOS2 branching implementation (graph from Nemhauser, Ahmed and V. '10)



■ Weak Integer   ■ SOS2 Branching   ■ Mystery Integer

# **Formulation Step 1: Encoding Alternatives**



# Formulation for Discrete Alternatives

$$\sum_{i=1}^n \lambda_i = 1$$

$$\sum_{i=1}^n b^i \lambda_i = y$$

$$\lambda \in \mathbb{R}_+^n$$

$$y \in \{0, 1\}^{\log_2 n}$$

$$\{b^i\}_{i=1}^n = \{0, 1\}^{\log_2 n}$$

- Li and Lu 2009, Adams and Henry 2011, V. and Nemhauser 2008.
- Sommer, TIMS 1972.
- Log = Binary Encoding
- Other choices of  $\{b^i\}_{i=1}^n$  lead to standard and incremental formulations

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# Unary Encoding

$$\begin{pmatrix} 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$

$$\lambda = y,$$

$$\sum_{i=1}^8 \lambda_i = 1,$$

$$\lambda \in \mathbb{R}^8, y \in \{0, 1\}^8$$



$$\lambda_i = y_i$$

# Binary Encoding

$$\begin{pmatrix} 1 & 1 & 1 & 1 & 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 \end{pmatrix} \lambda = y, \quad \begin{aligned} \sum_{i=1}^8 \lambda_i &= 1, \\ \lambda &\in \mathbb{R}^8, y \in \{0, 1\}^3 \end{aligned}$$

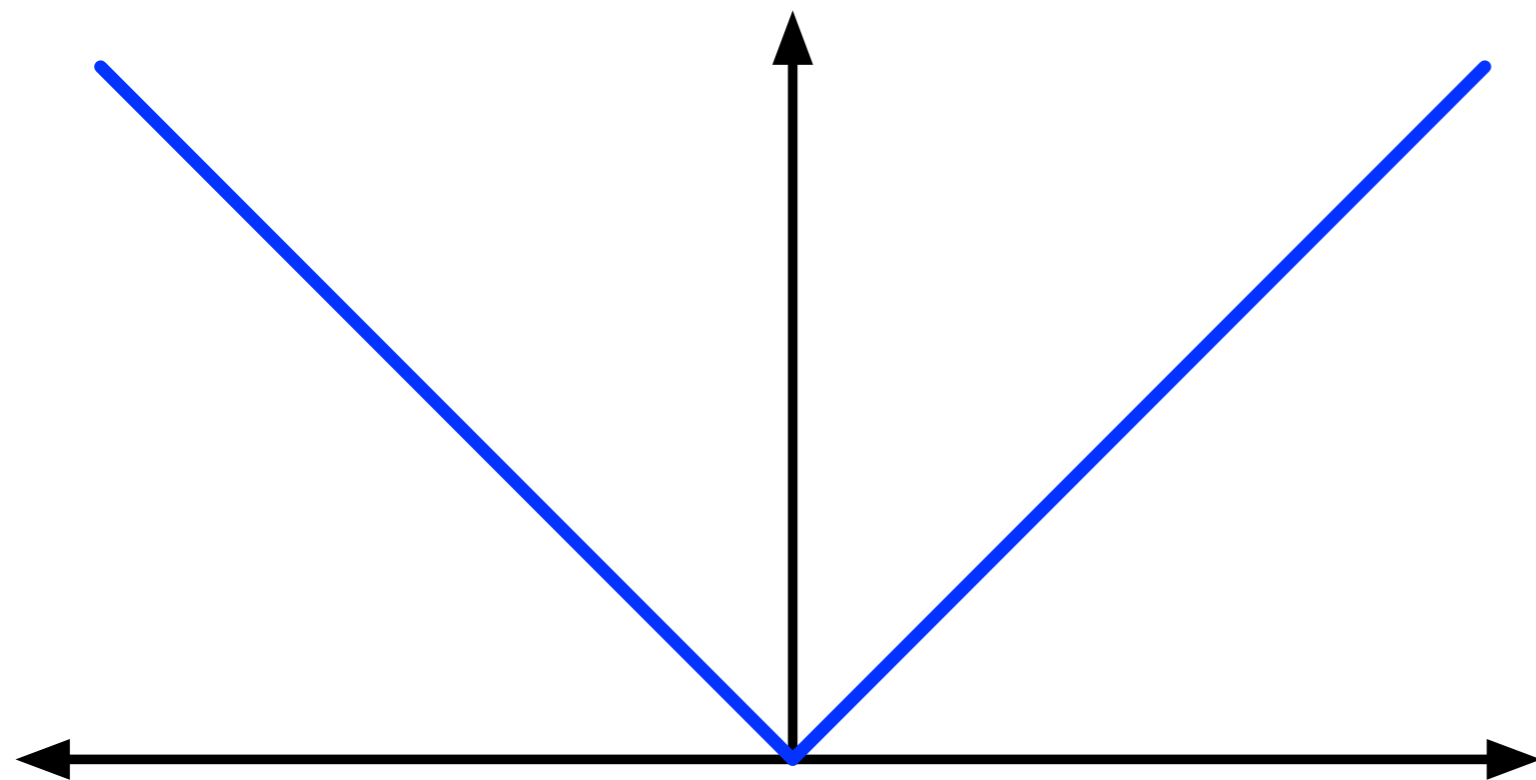
# Incremental Encoding

$$\begin{pmatrix} 0 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ 0 & 0 & 1 & 1 & 1 & 1 & 1 & 1 \\ 0 & 0 & 0 & 1 & 1 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 \end{pmatrix} \lambda = y, \quad \begin{aligned} \sum_{i=1}^8 \lambda_i &= 1, \\ \lambda &\in \mathbb{R}^8, y \in \{0, 1\}^7 \end{aligned}$$



$$y_1 \geq y_2 \geq \dots \geq y_7$$

# Example: Unary Encoding



min

$|x|$

s.t.

$$\sum_{i=1}^n \lambda_i \alpha_i = x$$

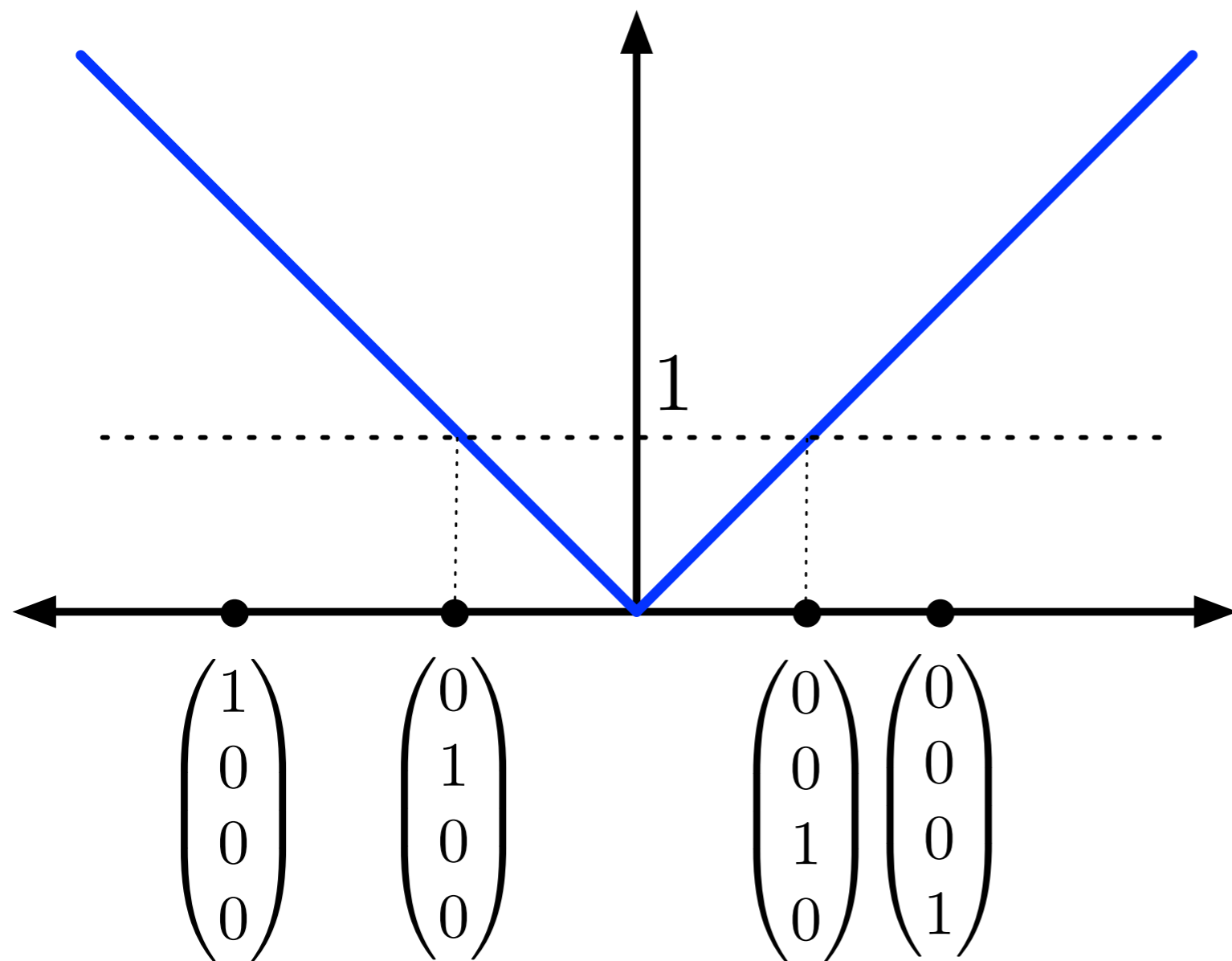
$$\sum_{i=1}^n \lambda_i = 1$$

$$\sum_{i=1}^n b^i \lambda_i = y$$

$$\lambda \in \mathbb{R}_+^n$$

$$y \in \{0, 1\}^m$$

# Example: Unary Encoding



min  $|x|$

s.t.

$$\sum_{i=1}^n \lambda_i \alpha_i = x$$

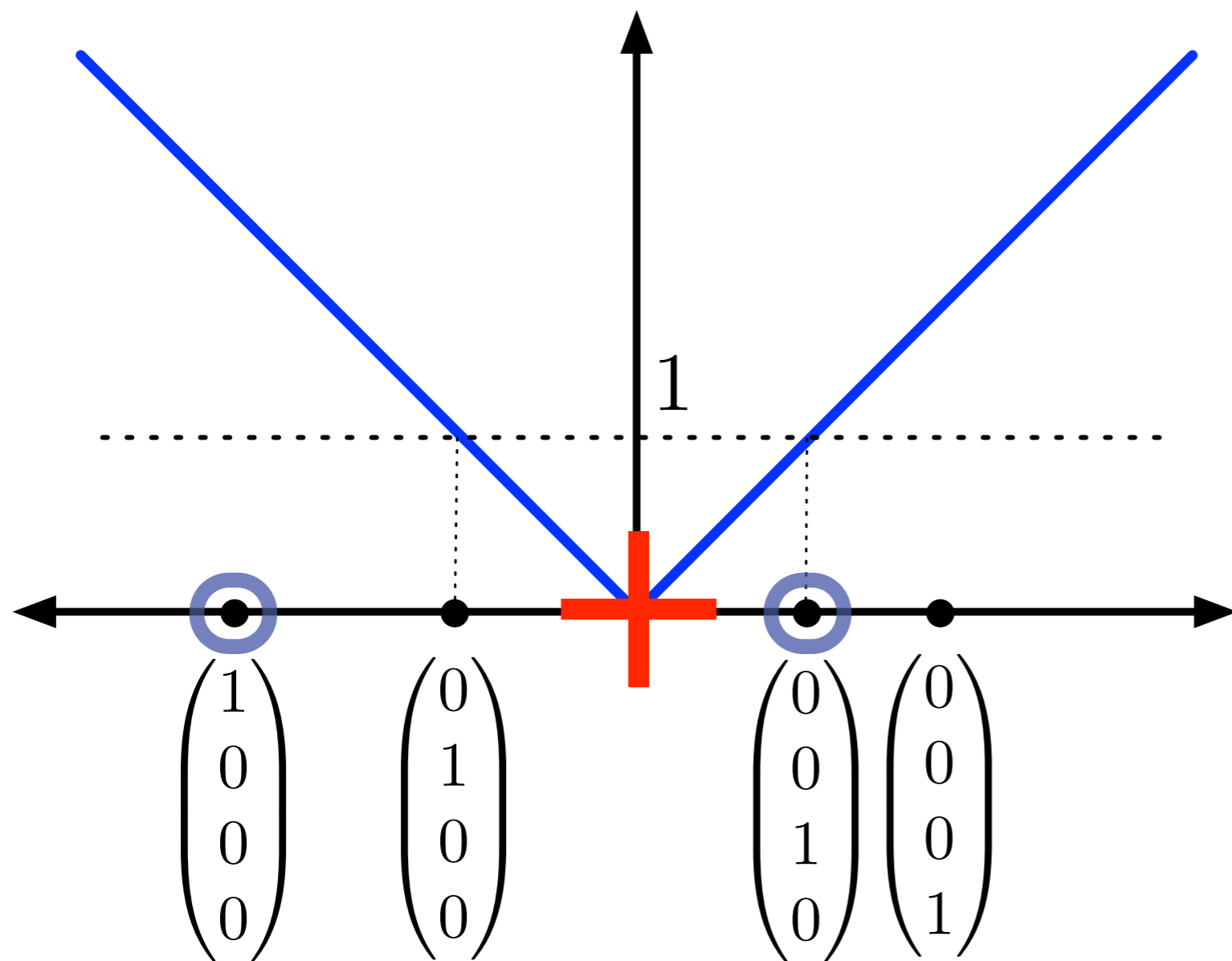
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# Example: Unary Encoding



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s.t.

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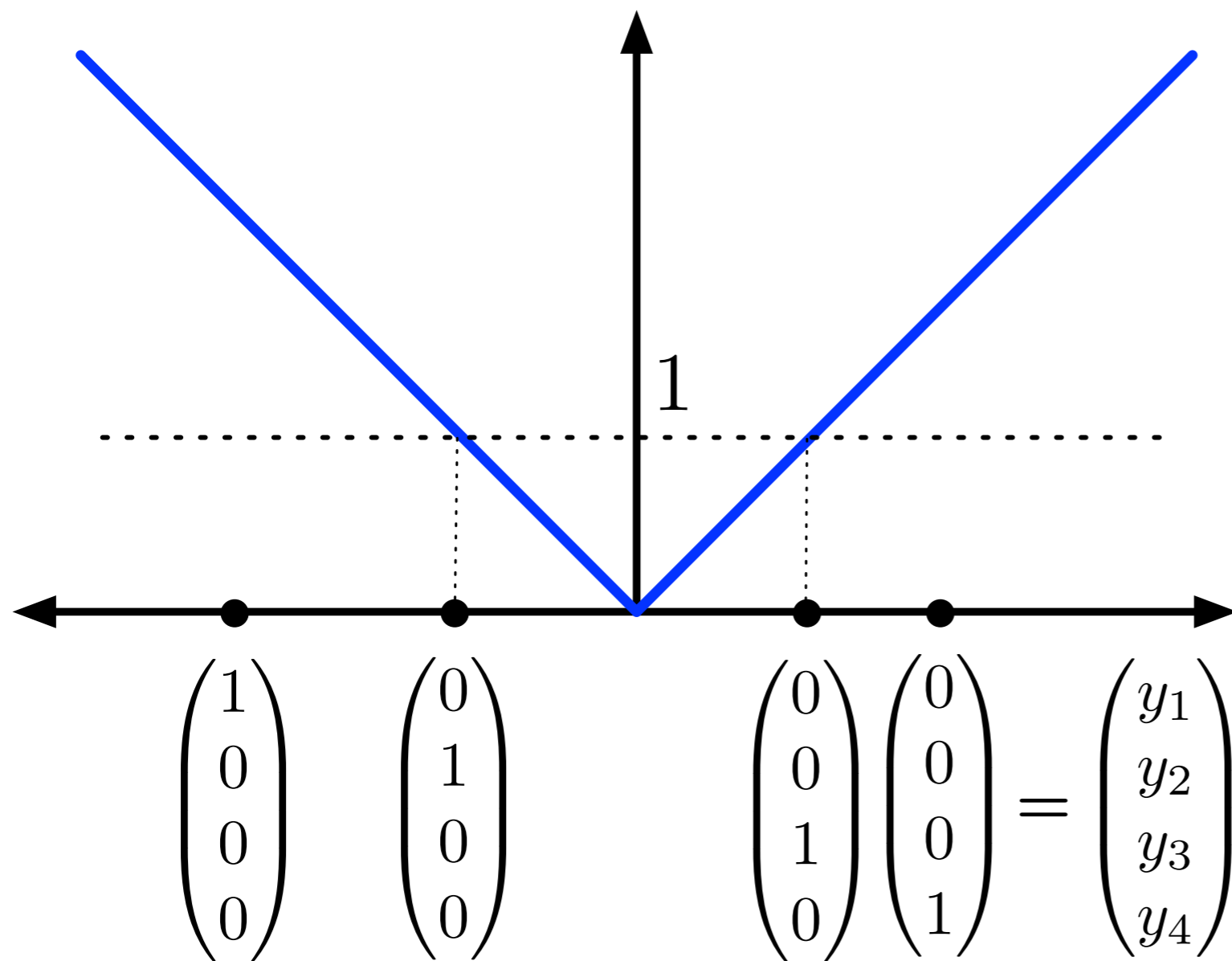
$$\sum_{i=1}^n b^i \lambda_i = y$$

$$\lambda \in \mathbb{R}_+^n$$

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# Example: Unary Encoding



min  $|x|$

s.t.

$$\sum_{i=1}^n \lambda_i \alpha_i = x$$

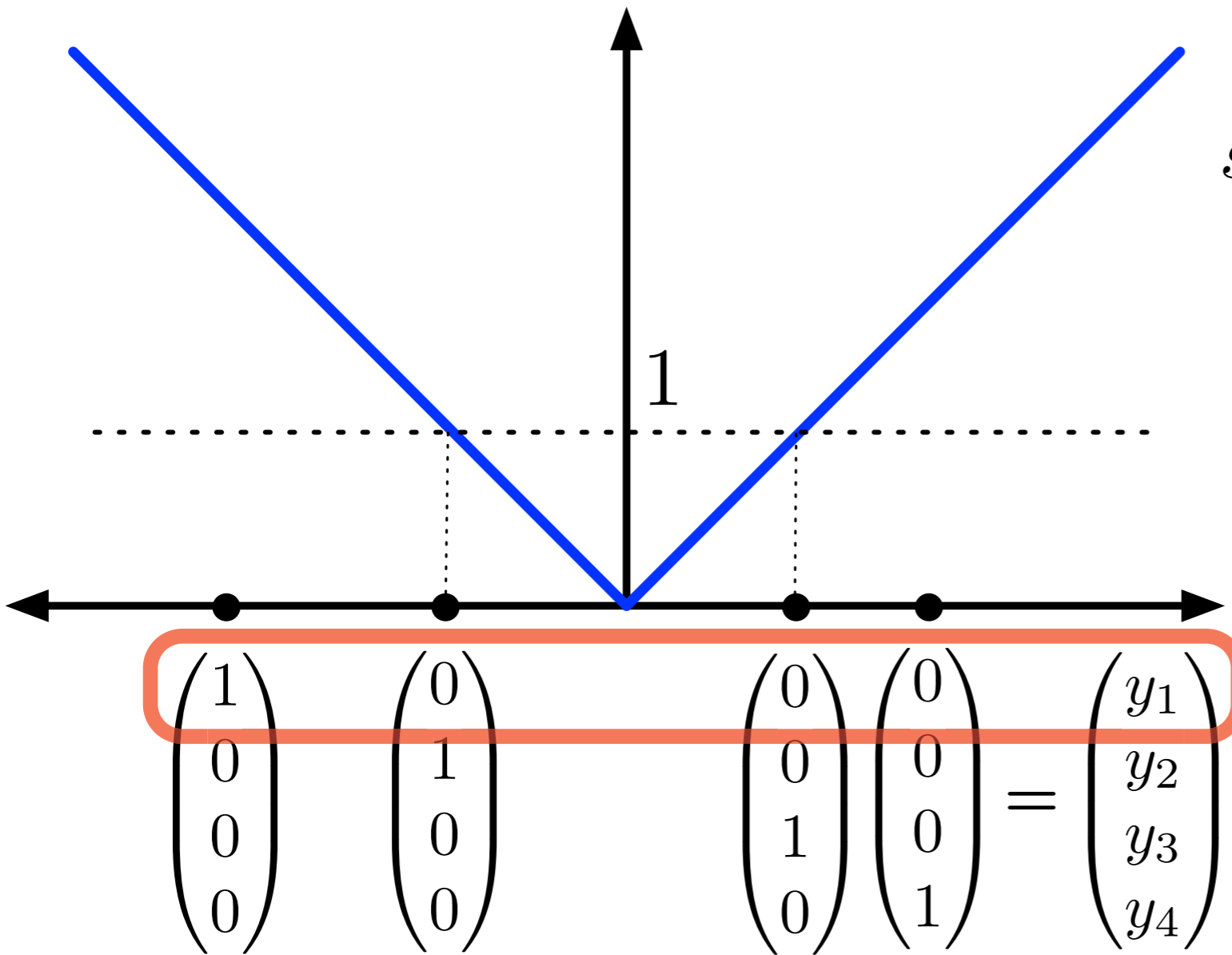
$$\sum_{i=1}^n \lambda_i = 1$$

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# Example: Unary Encoding



min  $|x|$   
 s.t.

$$\sum_{i=1}^n \lambda_i \alpha_i = x$$

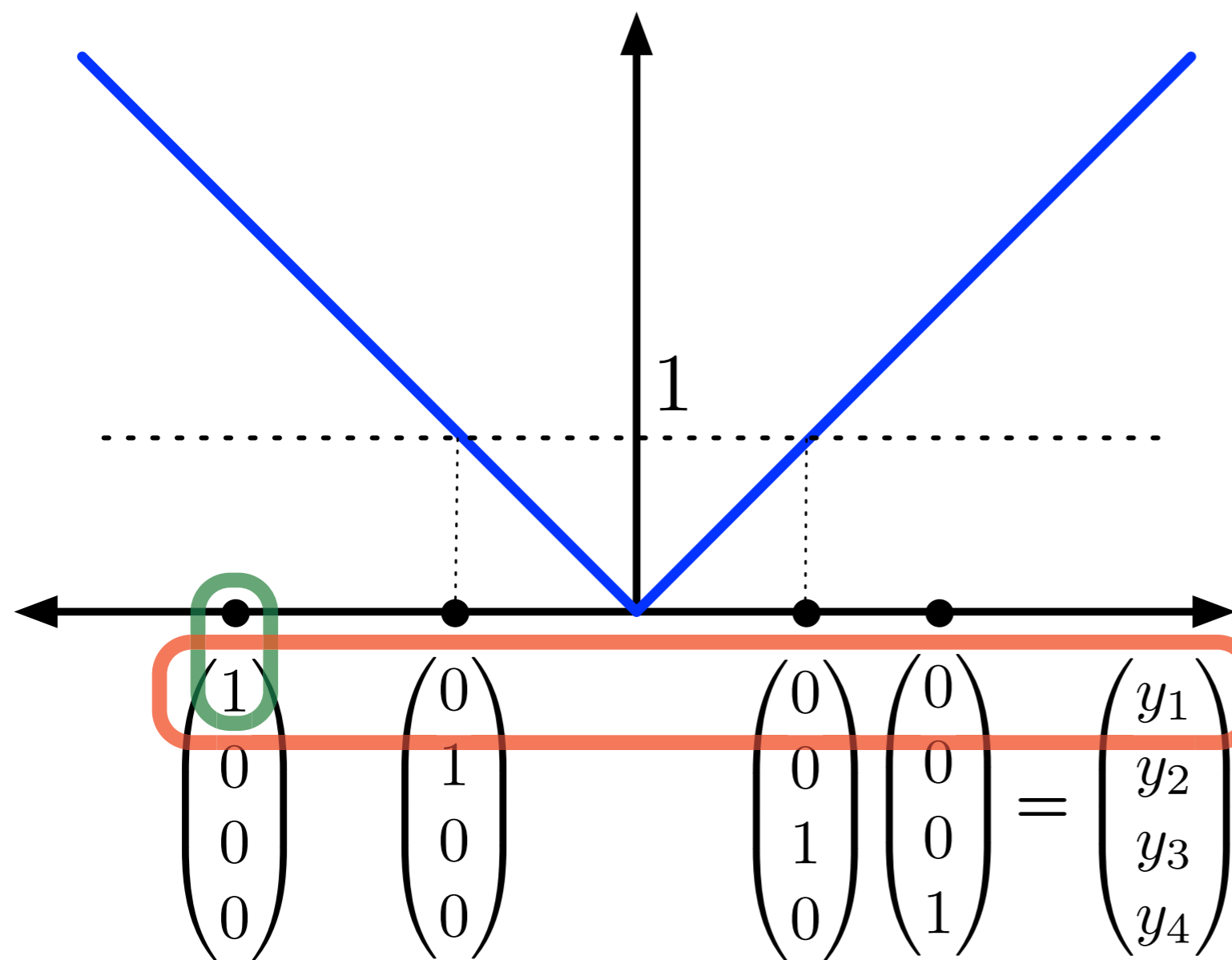
$$\sum_{i=1}^n \lambda_i = 1$$

$$\sum_{i=1}^n b^i \lambda_i = y$$

$\lambda \in \mathbb{R}_+^n$   
 $y \in \{0, 1\}^m$

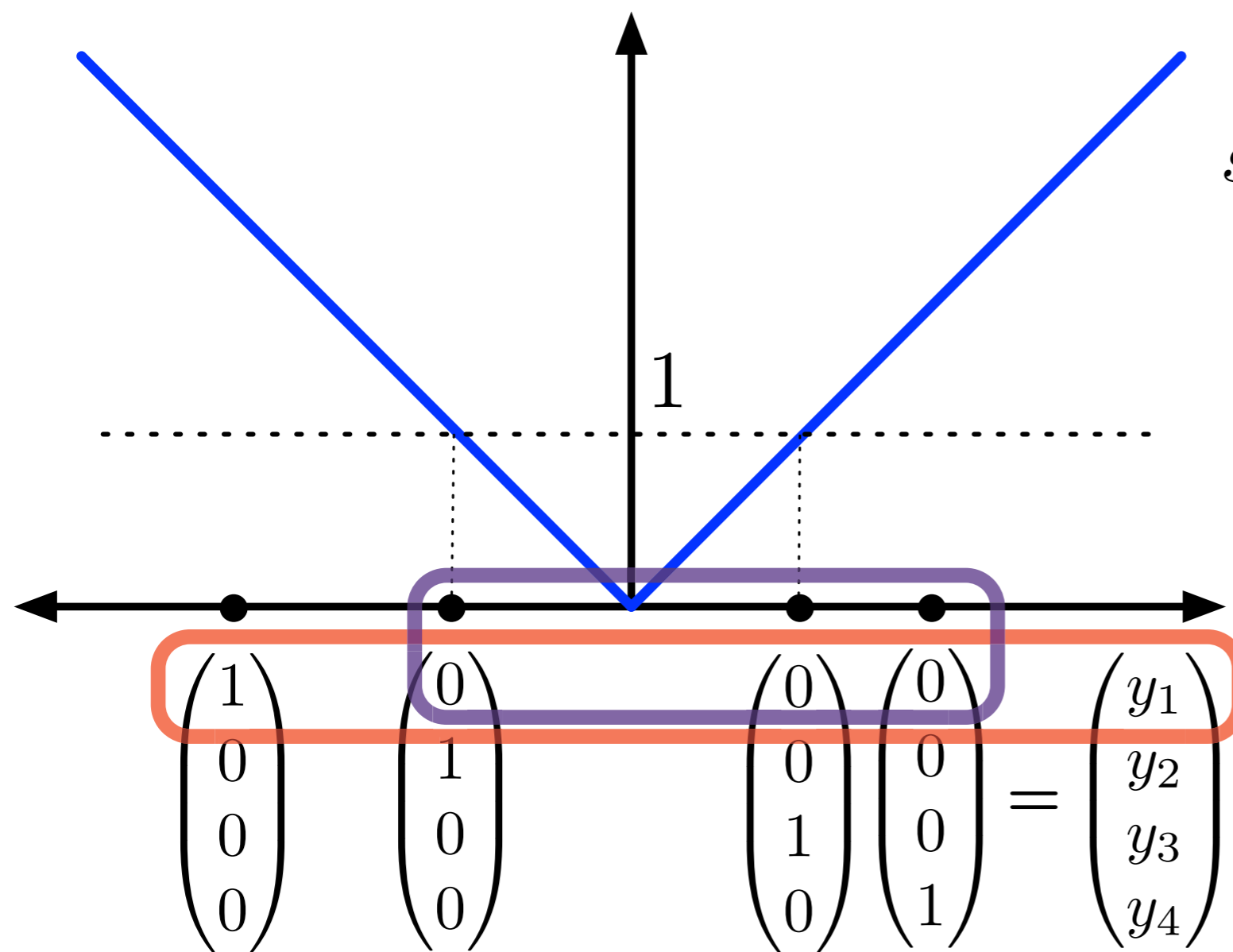
$y_1 =$

# Example: Unary Encoding



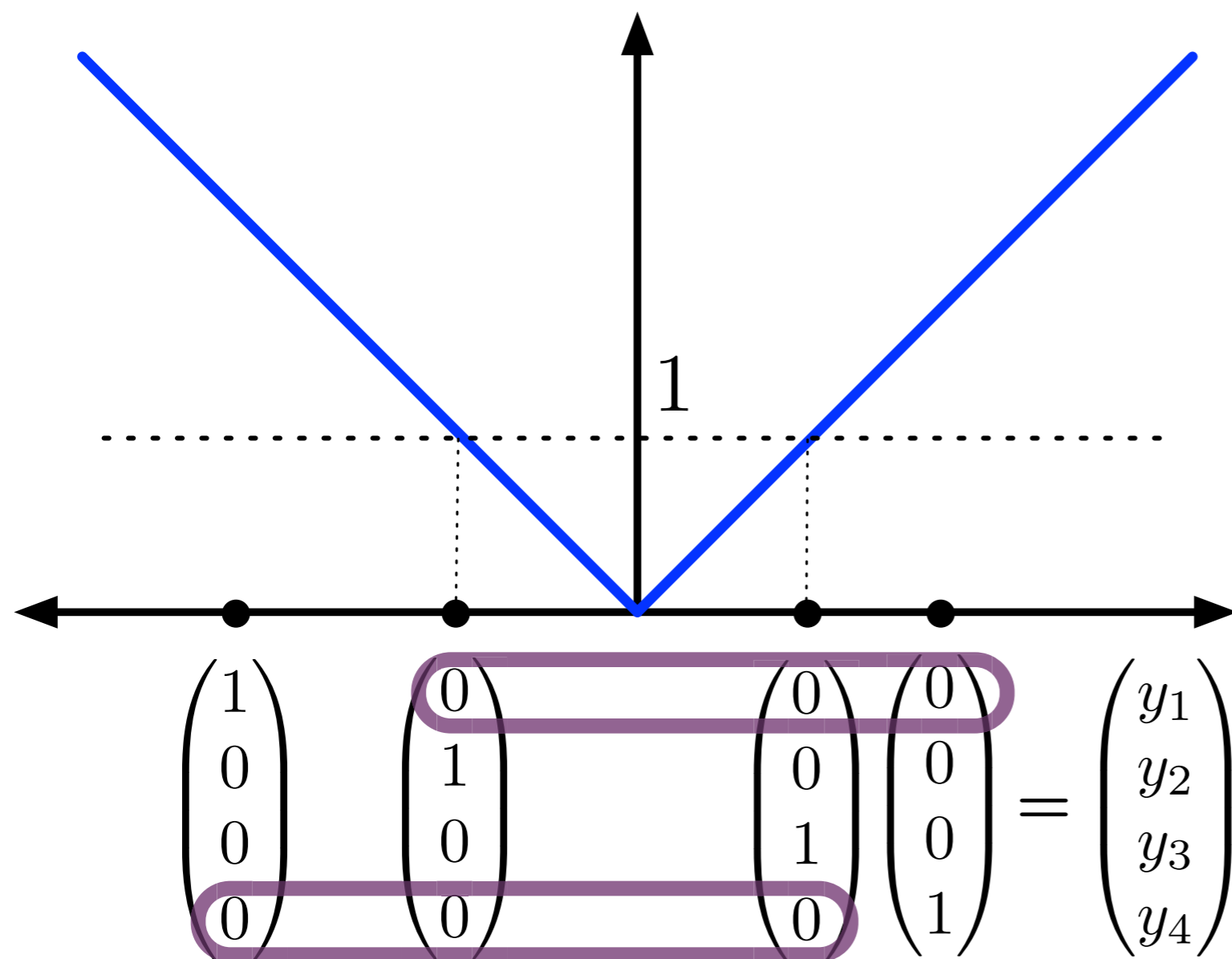
$$y_1 = 1$$

# Example: Unary Encoding



$$y_1 = 0$$

# Example: Unary Encoding



min

 $|x|$ 

s.t.

$$\sum_{i=1}^n \lambda_i \alpha_i = x$$

$$\sum_{i=1}^n \lambda_i = 1$$

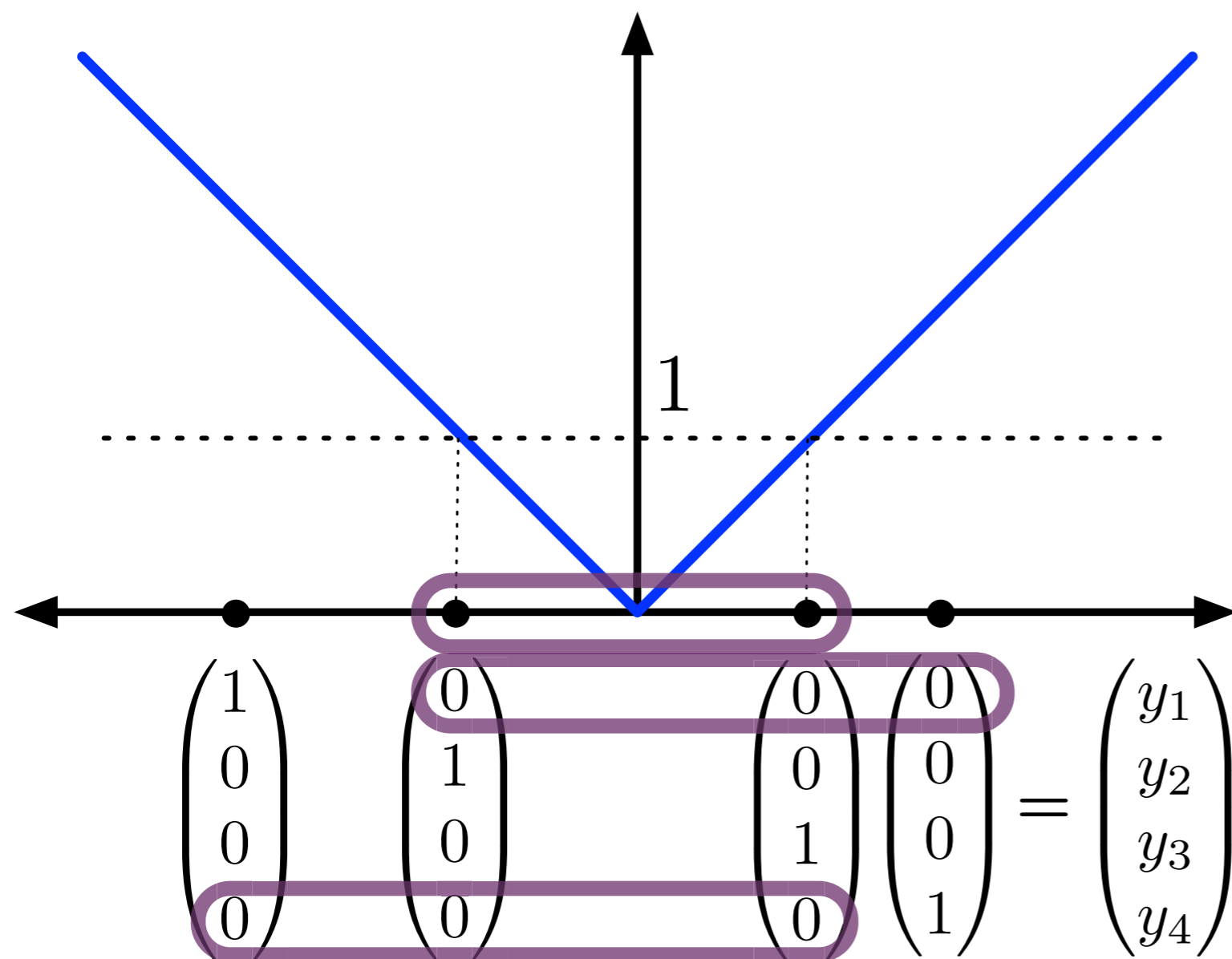
$$\sum_{i=1}^n b^i \lambda_i = y$$

$$\lambda \in \mathbb{R}_+^n$$

$$y \in \{0, 1\}^m$$

$$y_1 = y_4 = 0$$

# Example: Unary Encoding



min

 $|x|$ 

s.t.

$$\sum_{i=1}^n \lambda_i \alpha_i = x$$

$$\sum_{i=1}^n \lambda_i = 1$$

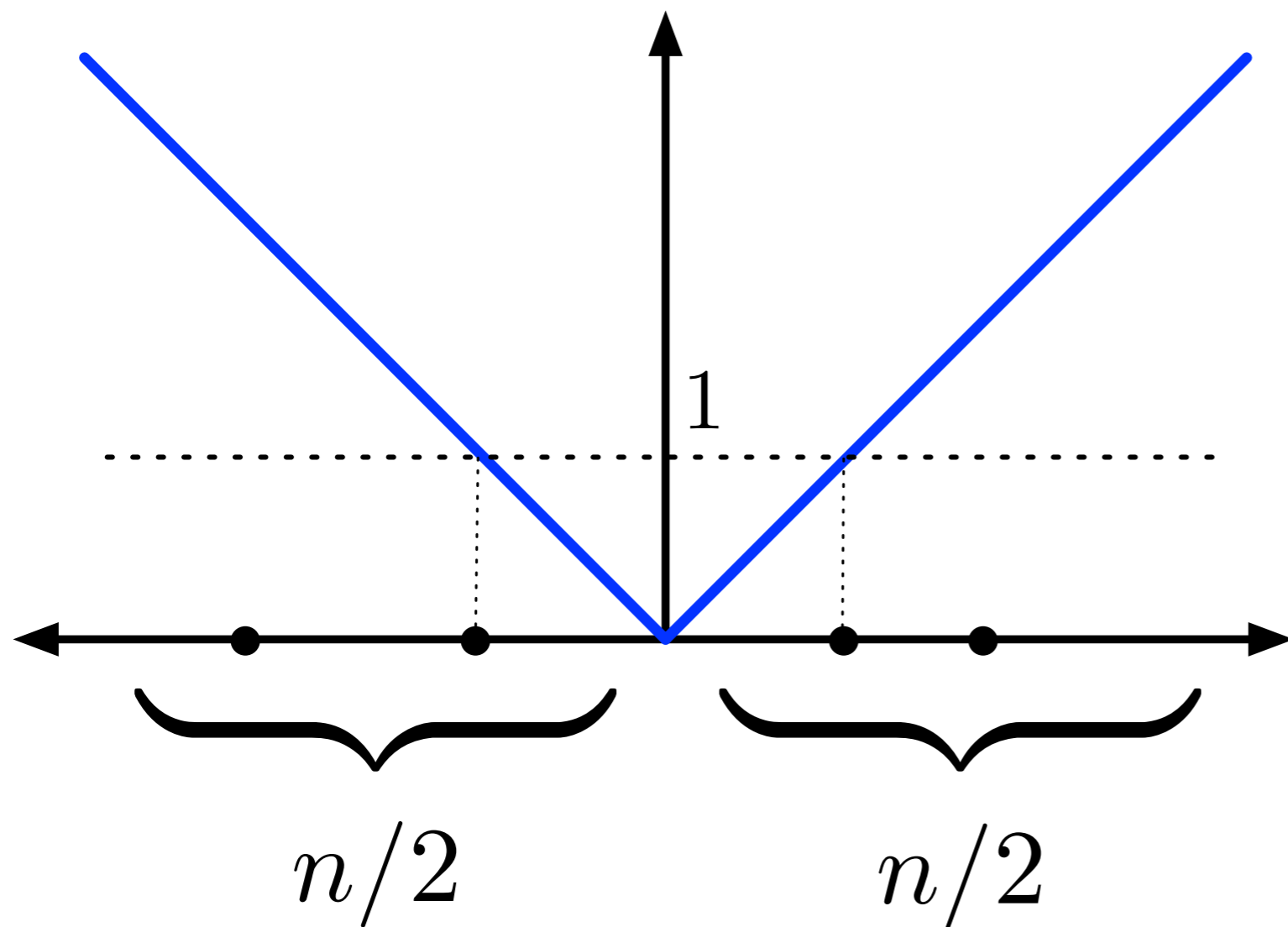
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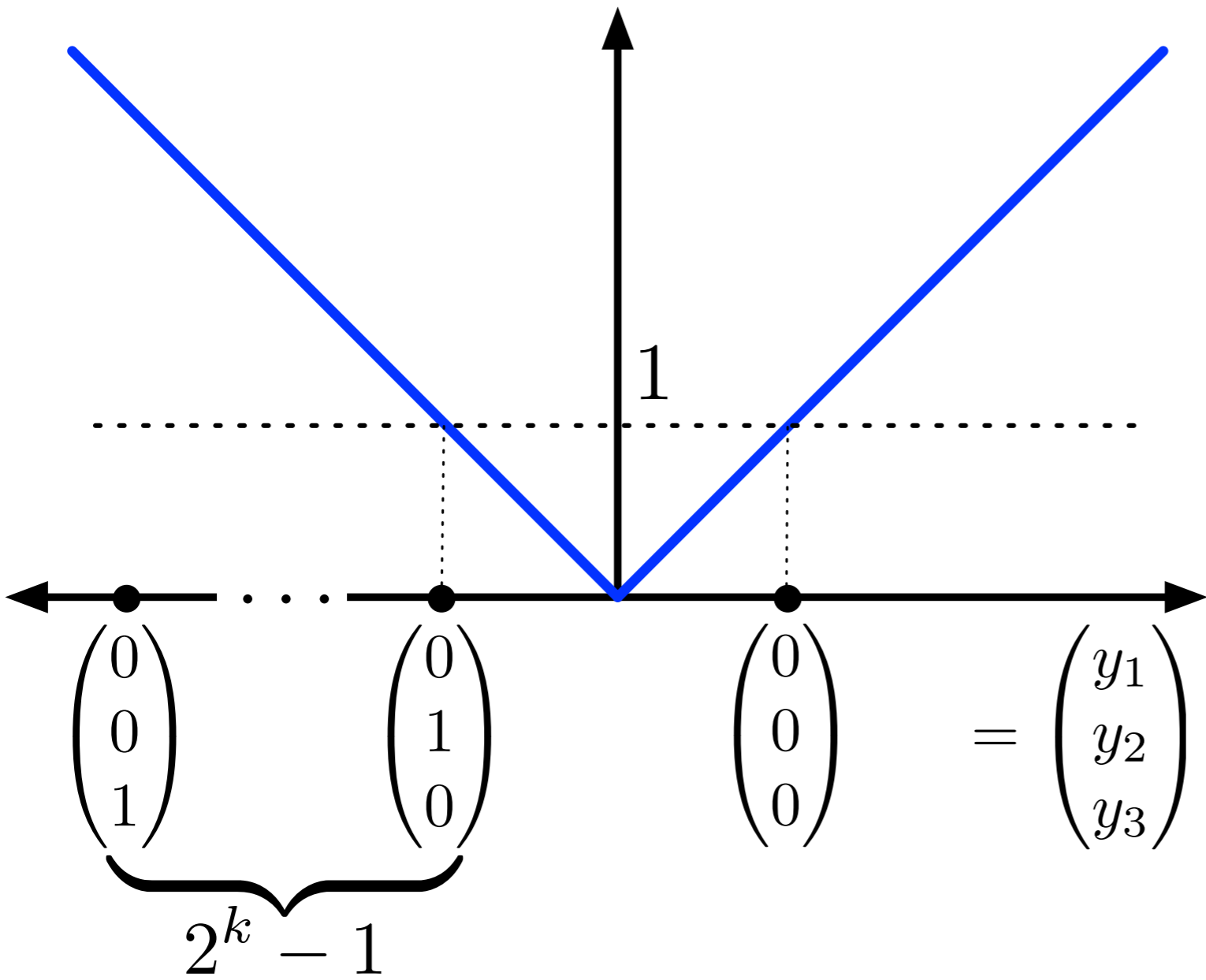
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$$y \in \{0, 1\}^m$$

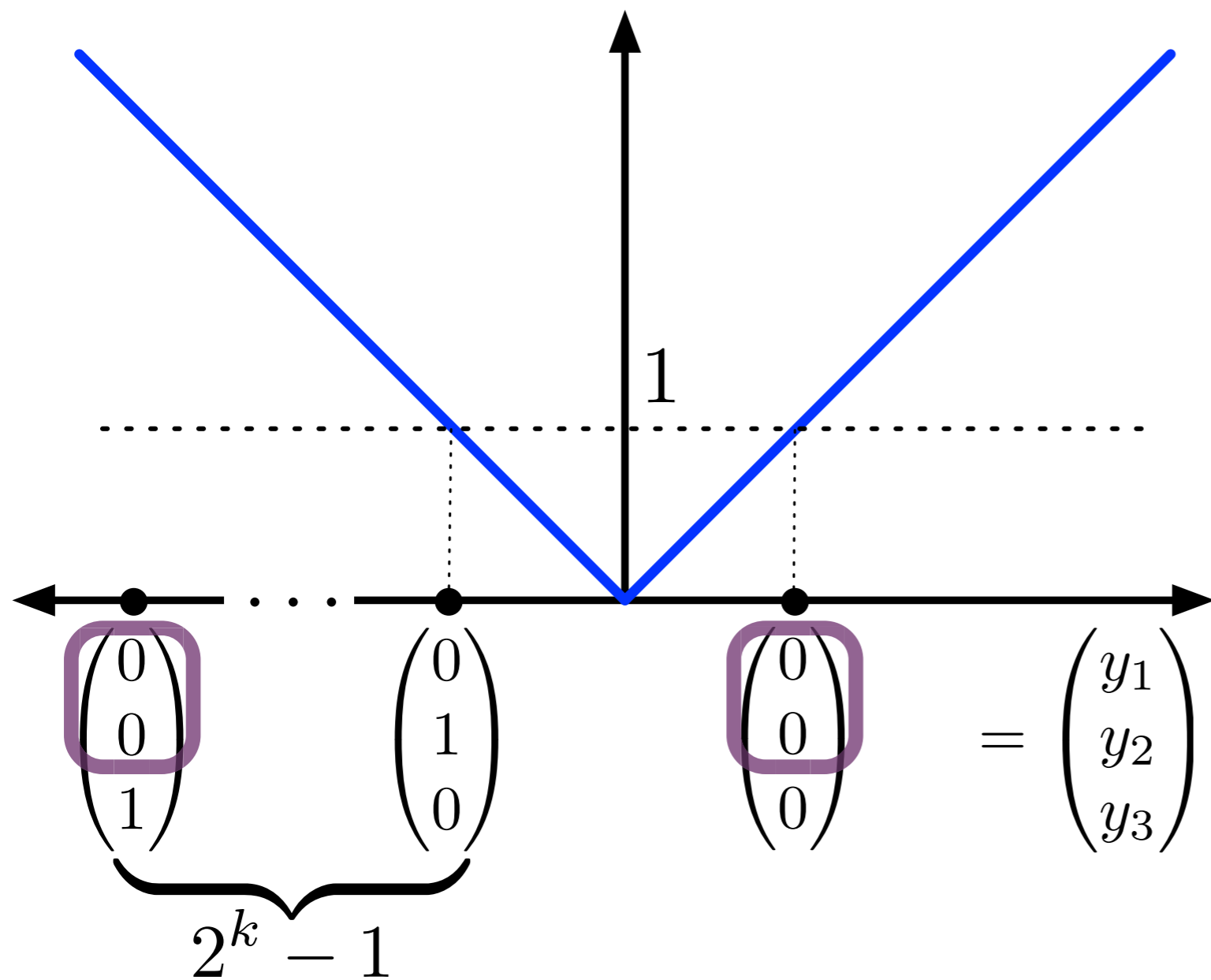
Need  $n/2$   
branches

# Example: Binary Encoding



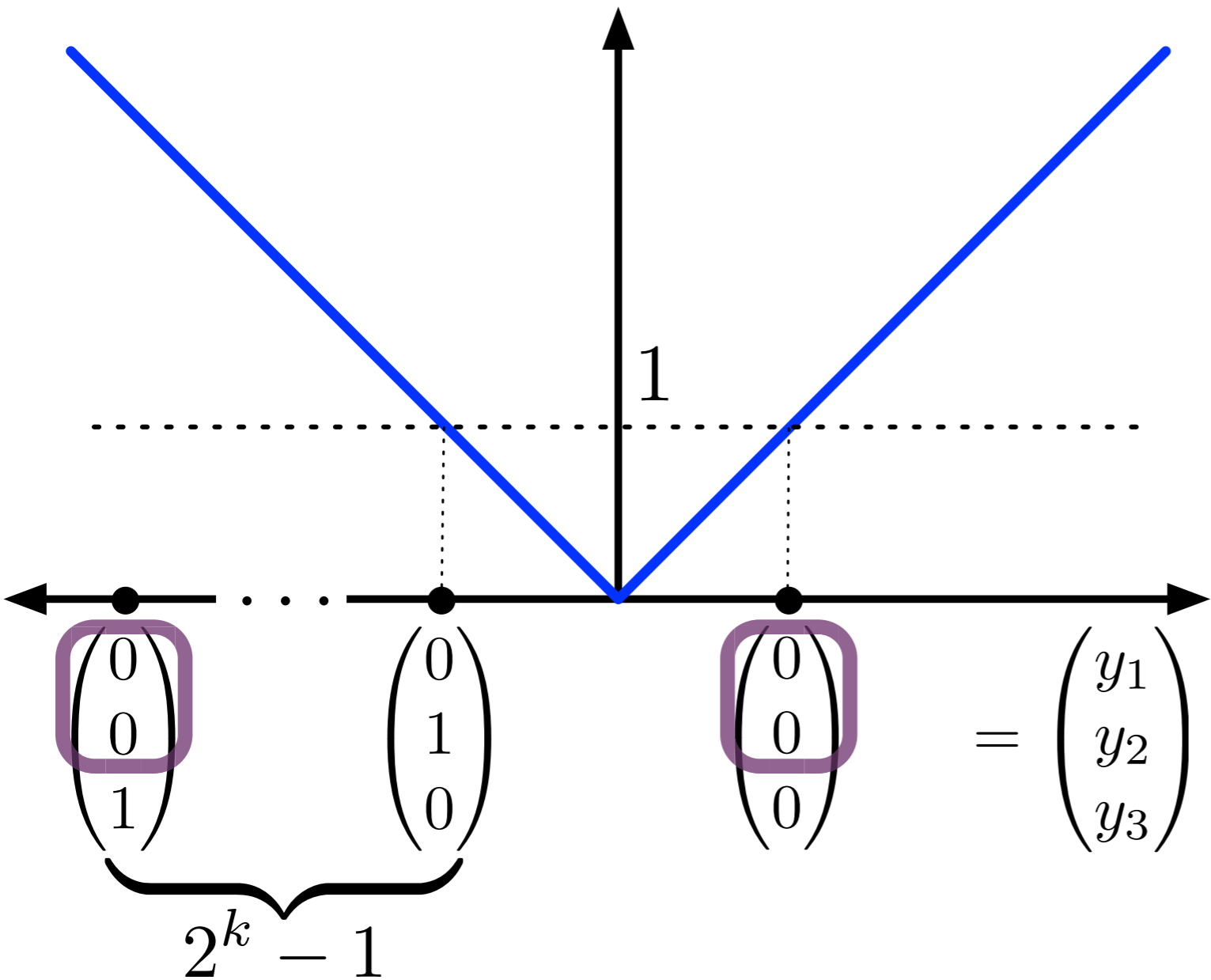


# Example: Binary Encoding



$$y_1 = y_2 = 0$$

# Example: Binary Encoding

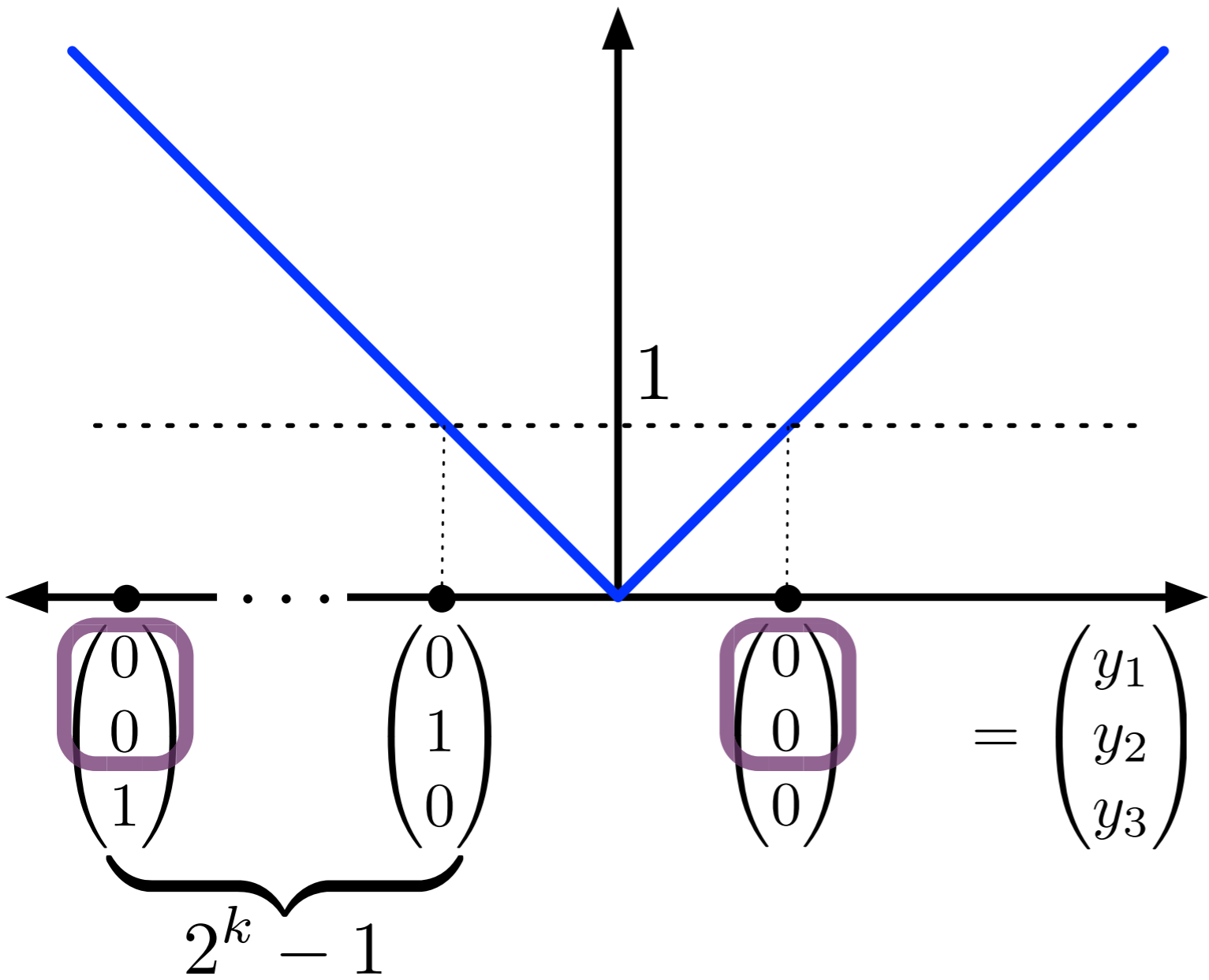


Best Bound = 0 unless:

$$y_i = 0 \quad \forall i$$

$$y_1 = y_2 = 0$$

# Example: Binary Encoding



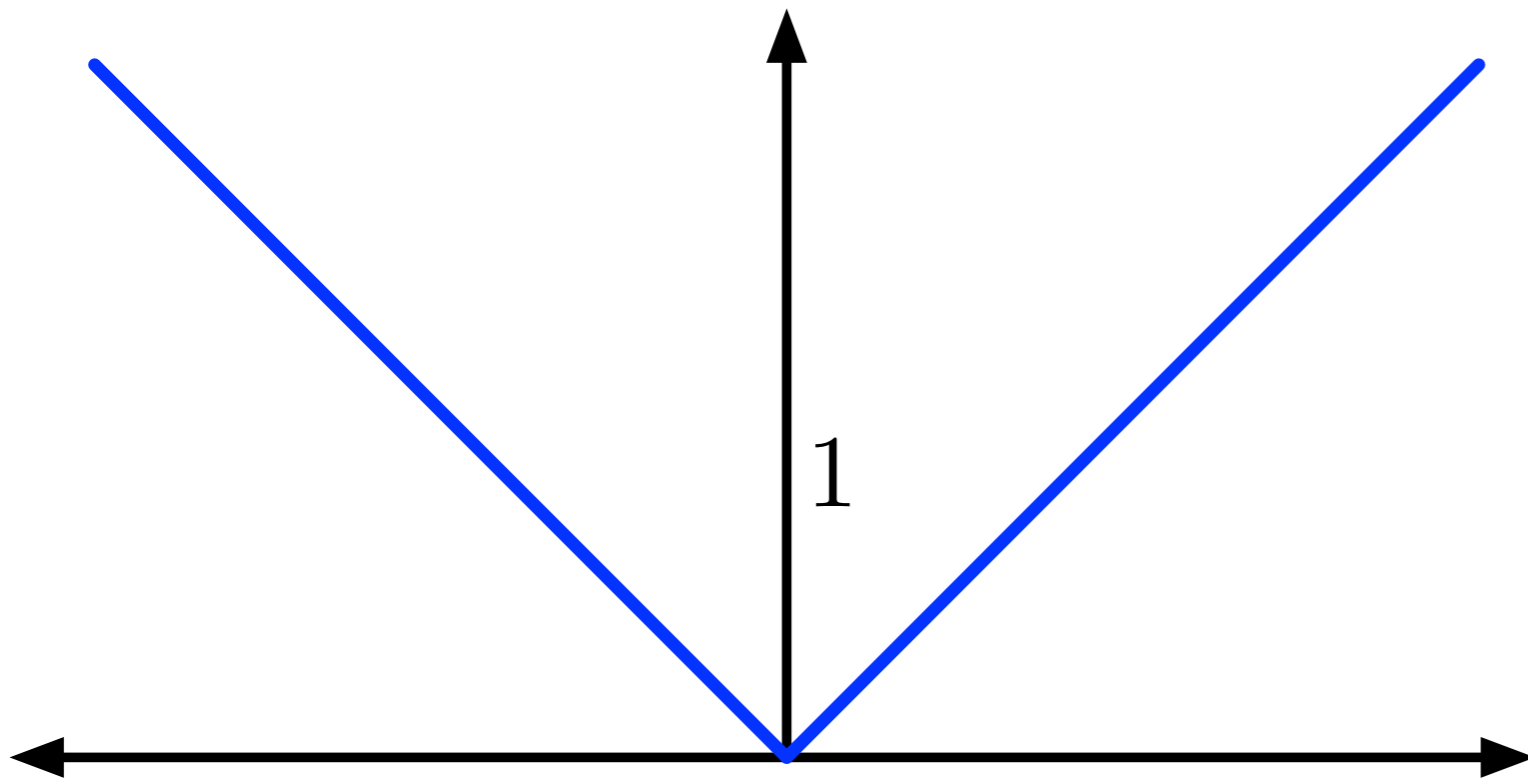
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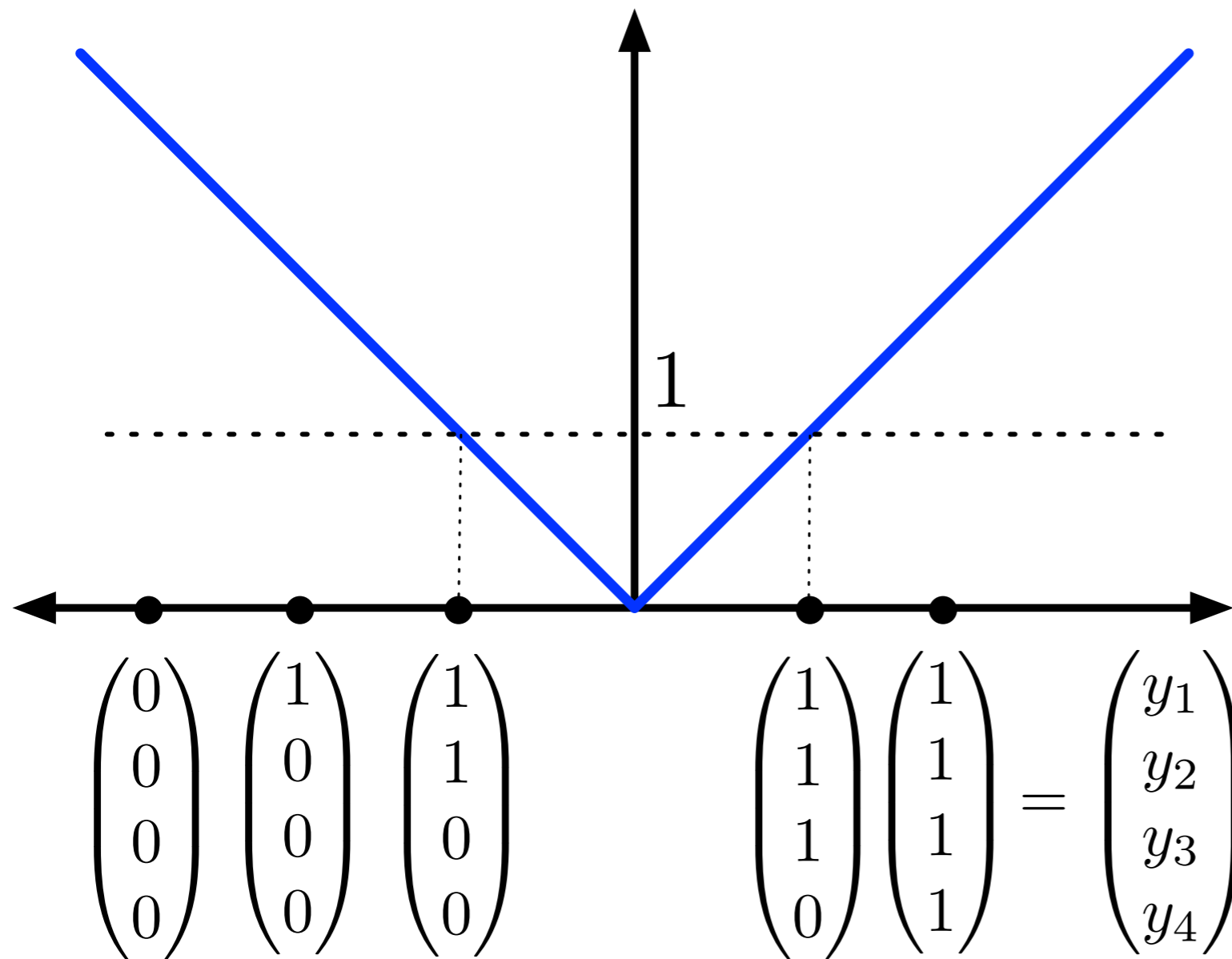
Need  $k = \log_2 n$  branches

$$y_1 = y_2 = 0$$

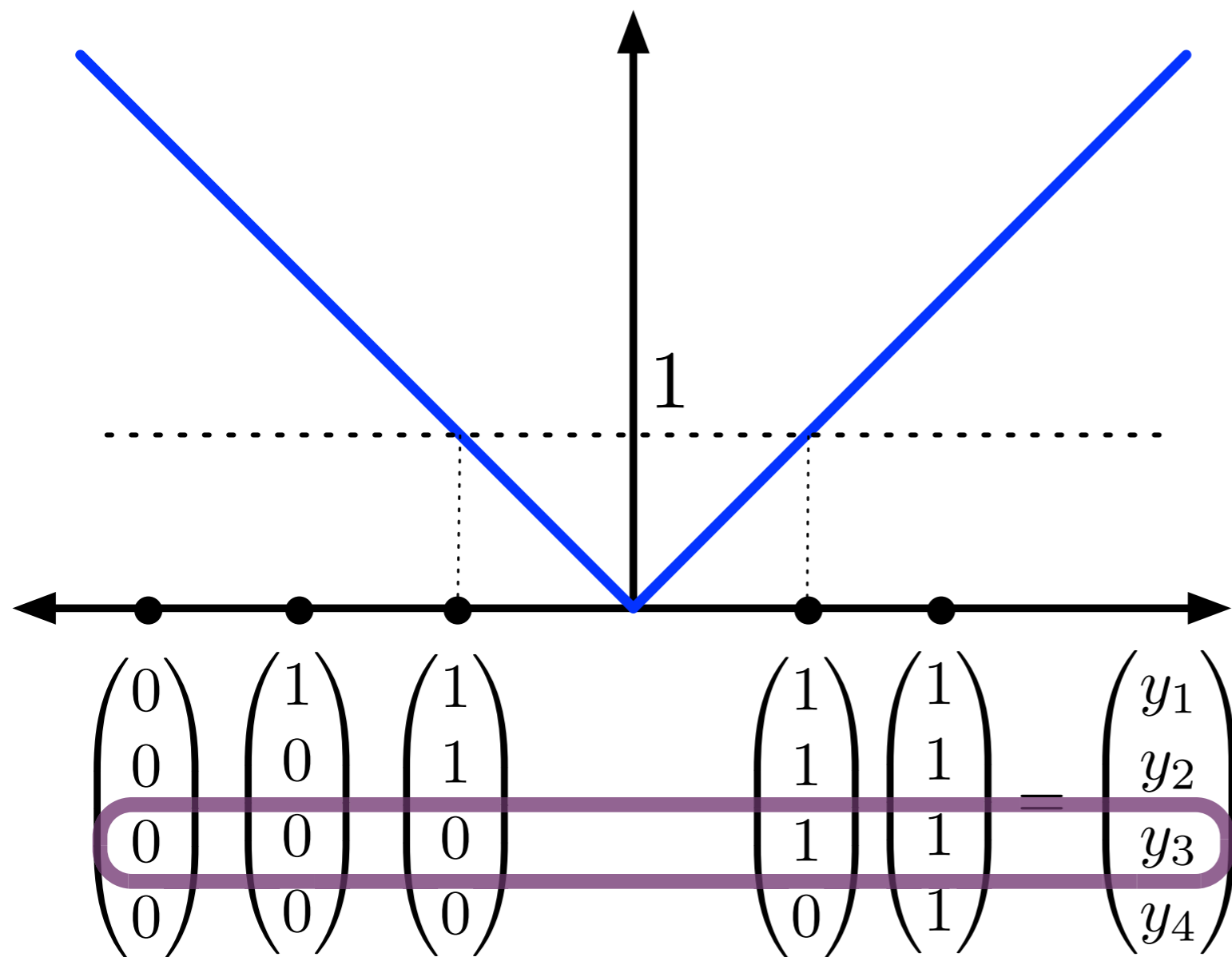
# Example: Incremental Encoding



# Example: Incremental Encoding

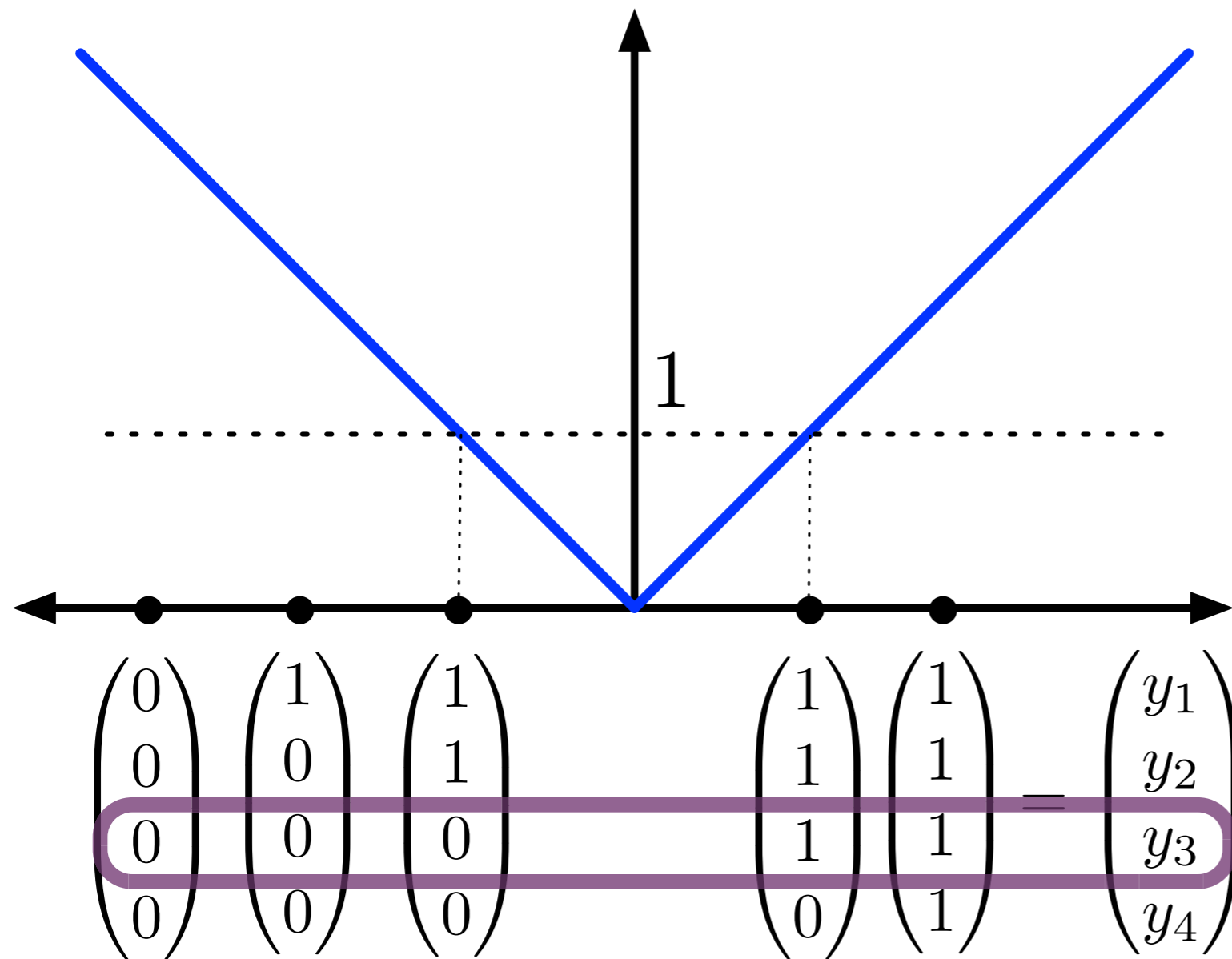


# Example: Incremental Encoding



$$y_3 = 1 \vee y_3 = 0$$

# Example: Incremental Encoding



Best Bound = 1 if:  
 $y_{i^*} = 0 \vee y_{i^*} = 1$

Only need  
 1 branch!

$$y_3 = 1 \vee y_3 = 0$$

# Induced Constraint Branching

Incremental

$$\begin{pmatrix} 0 & 1 & 1 & 1 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 \end{pmatrix} \lambda = y$$

Binary

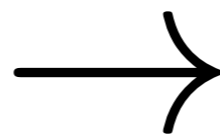
$$\begin{pmatrix} 0 & 0 & 1 & 1 \\ 0 & 1 & 0 & 1 \end{pmatrix} \lambda = y$$



# Induced Constraint Branching

Incremental

$$\begin{pmatrix} 0 & 1 & 1 & 1 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 \end{pmatrix} \lambda = y$$



SOS1 Branching

$$\lambda_1 = \lambda_2 = 0$$

*or*

$$\lambda_3 = \lambda_4 = 0$$

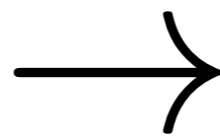
Binary

$$\begin{pmatrix} 0 & 0 & 1 & 1 \\ 0 & 1 & 0 & 1 \end{pmatrix} \lambda = y$$

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Incremental

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SOS1 Branching

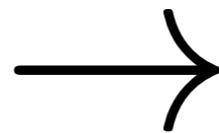
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Binary

$$\begin{pmatrix} 0 & 0 & 1 & 1 \\ 0 & 1 & 0 & 1 \end{pmatrix} \lambda = y$$



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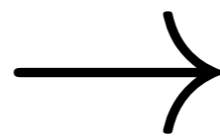
*or*

$$\lambda_3 = \lambda_4 = 0$$

# Induced Constraint Branching

Incremental

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SOS1 Branching

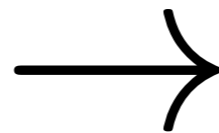
$$\lambda_1 = \lambda_2 = 0$$

*or*

$$\lambda_3 = \lambda_4 = 0$$

Binary

$$\begin{pmatrix} 0 & 0 & 1 & 1 \\ 0 & 1 & 0 & 1 \end{pmatrix} \lambda = y$$



Odd/Even Branching

$$\lambda_1 = \lambda_3 = 0$$

*or*

$$\lambda_2 = \lambda_4 = 0$$

# **Formulation Step 2: Combining with Strong Formulation**

# Long Lost Integral Formulation

$\{P^i\}_{i=1}^n$  polytopes

$$x \in \bigcup_{i=1}^n P^i \Leftrightarrow$$

$$\sum_{i=1}^n \sum_{v \in \text{ext}(P^i)} v \lambda_v^i = x$$

$$\sum_{v \in \text{ext}(P^i)} \lambda_v^i = y_i$$

$$\sum_{i=1}^n y_i = 1$$

$$y \in \{0, 1\}^n, \lambda_v^i \geq 0$$

Also for general polyhedra  
with common recession cones.

- Jeroslow and Lowe 1984.

# Combining with Alternative Encoding

$\{P^i\}_{i=1}^n$  polytopes

$$x \in \bigcup_{i=1}^n P^i \Leftrightarrow$$

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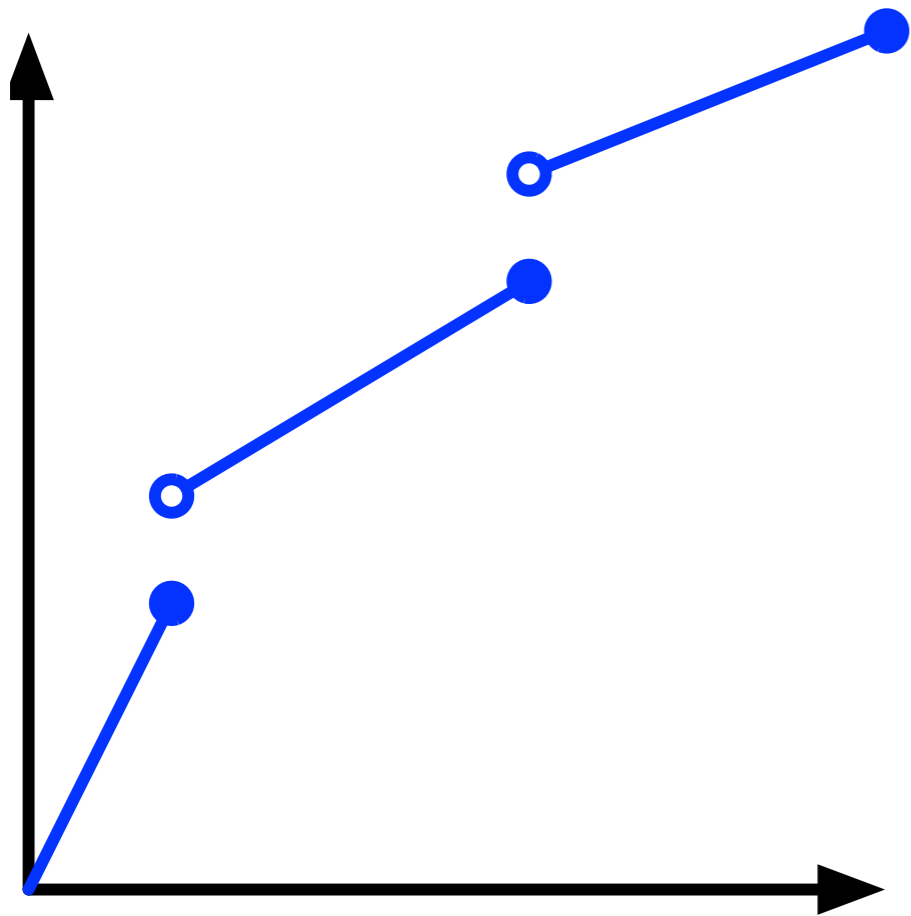
$$y \in \{0, 1\}^m, \lambda_v^i \geq 0$$

Also for general polyhedra  
with common recession cones.

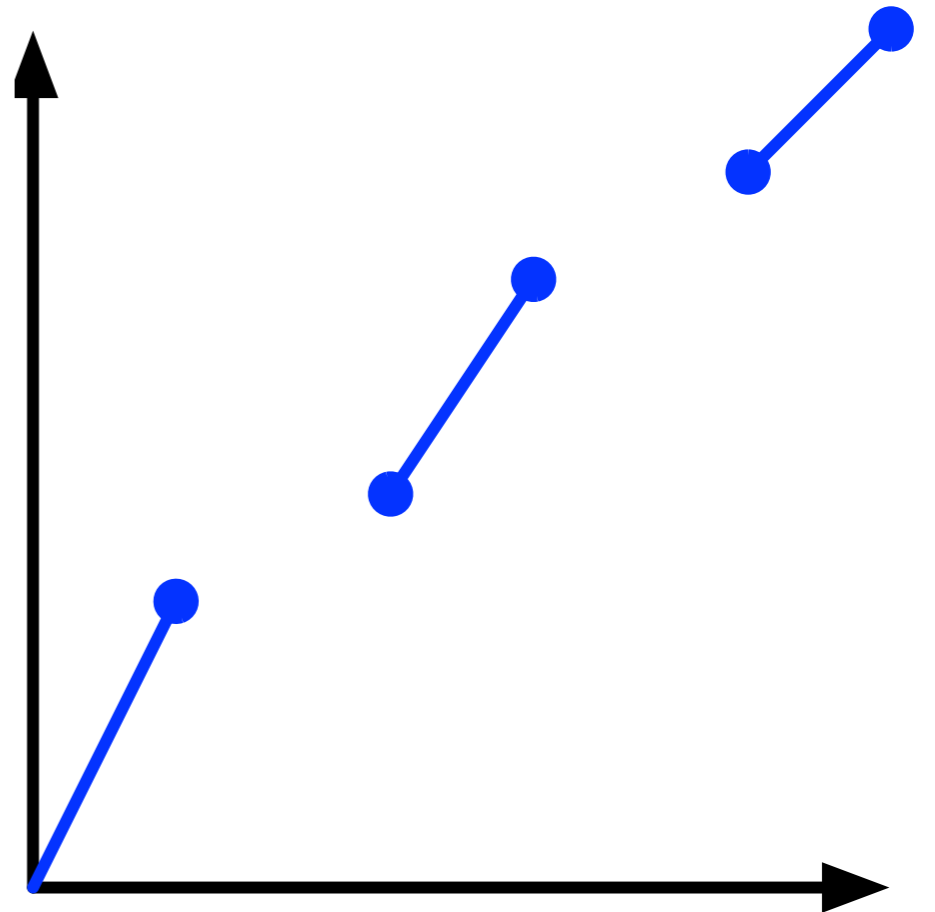
- V., Ahmed and Nemhauser 2010; V. 2012.



# Univariate Transportation Problems



Discontinuous  
Piecewise Linear

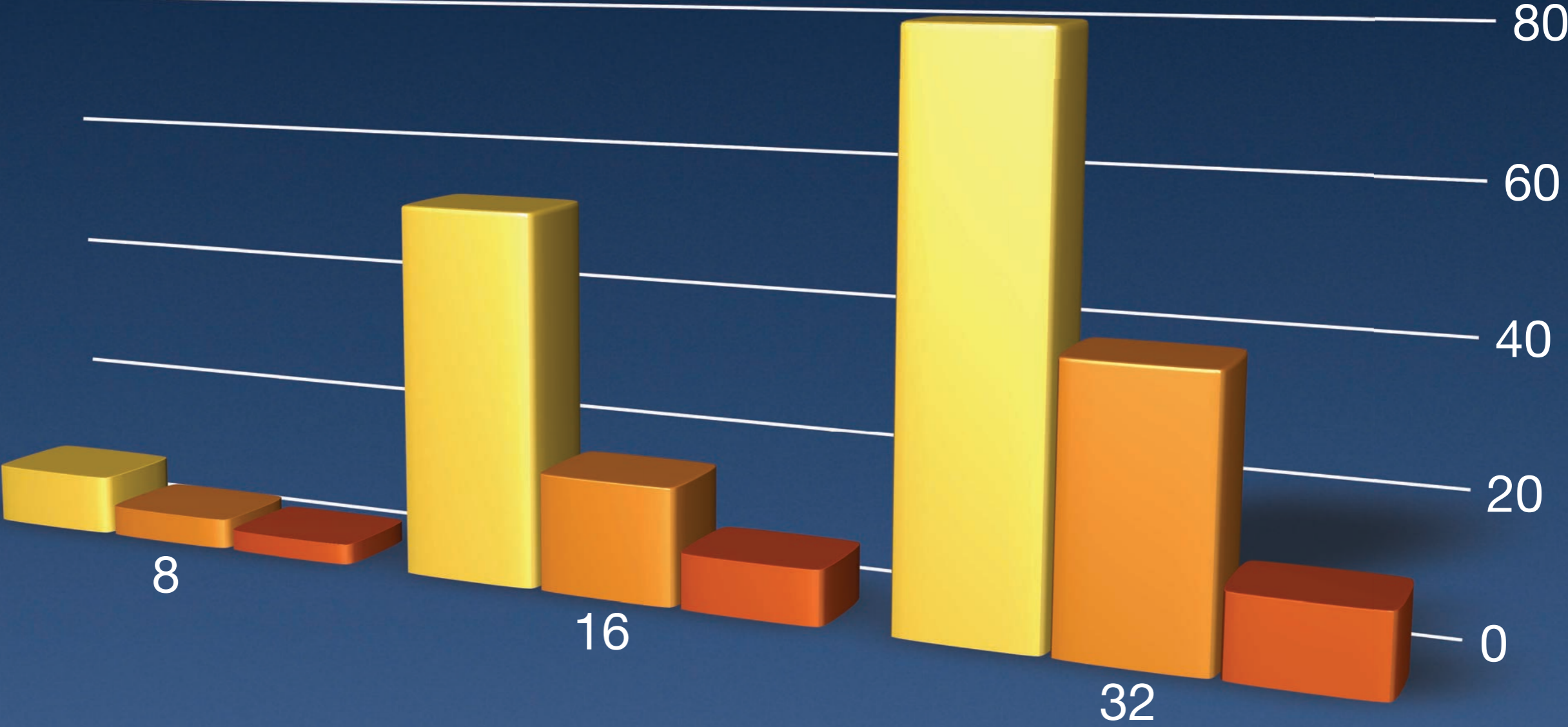


Disc. PWL  
+  
“Semicontinuous”

# Piecewise Linear

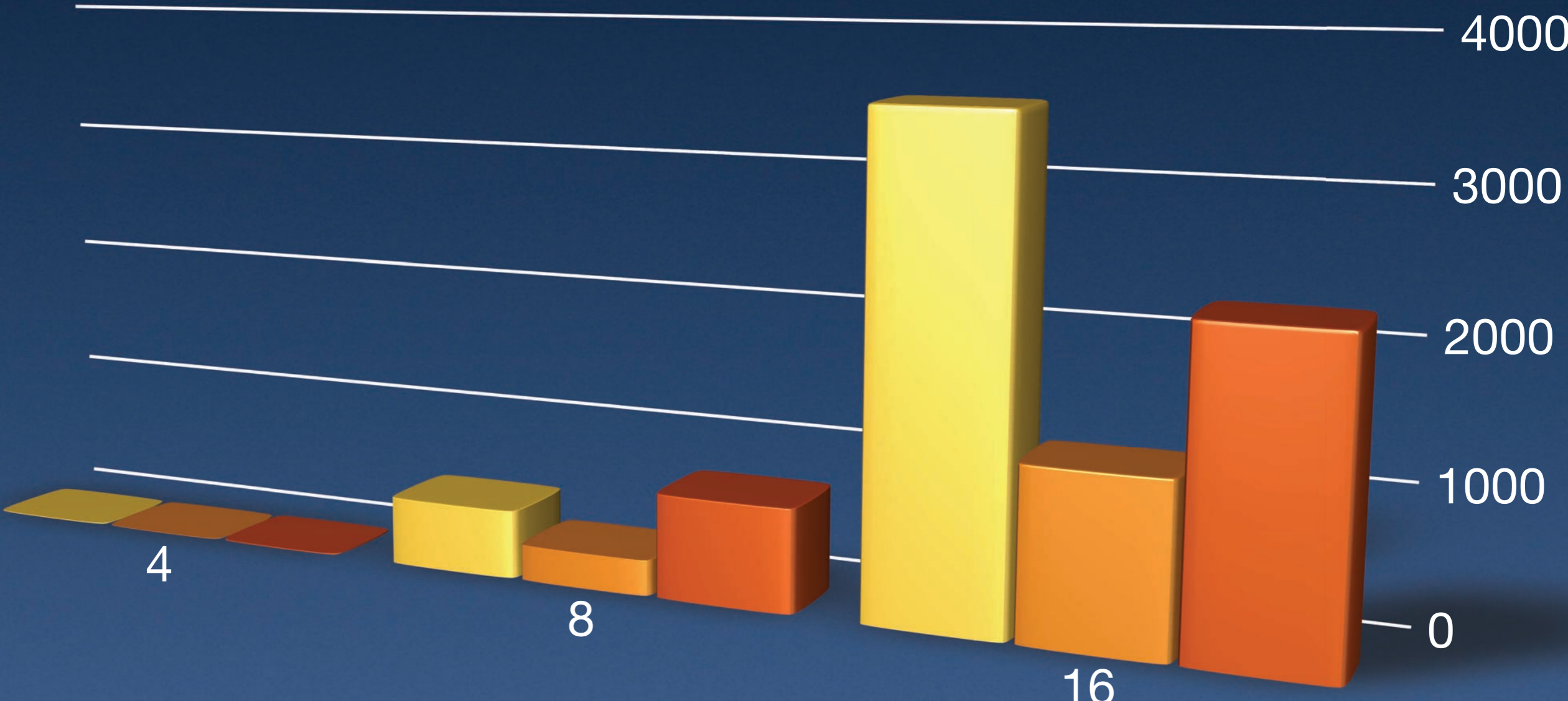
■ Unary    ■ Incremental / Strong Integer    ■ Binary / Mystery Integer

# Piecewise Linear



■ Unary    ■ Incremental / Strong Integer    ■ Binary / Mystery Integer

# Piecewise Linear + Semi Continuous



■ Unary    ■ Incremental / Strong Integer    ■ Binary / Mystery Integer

# Summary, Extensions and More.

- Effective formulations: Encode and Formulate
  - Best encoding? Why not try a few.
  - Clever combination of encodings can be useful (e.g. V. and Nemhauser 2008 for multivariate piecewise linear functions)
- Smaller formulations for shared vertex case
  - Need encodings with special structure.

# More Information

# More Information

- Survey: V., “MIP Formulation Techniques”:
  - [http://www.optimization-online.org/DB\\_HTML/2012/07/3539.html](http://www.optimization-online.org/DB_HTML/2012/07/3539.html).

# More Information

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- Next year: automatic formulations for JUMP
  - Julia based modeling language:
    - As simple as AMPL + “faster” than C++
    - Solver independent call-backs and more!
  - JUMP/Julia tutorial in January