

# Piecewise-Linear Approximation: Multidimensional

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Motivation

CC Model

DCC Model

SOS2 Model

# Summary

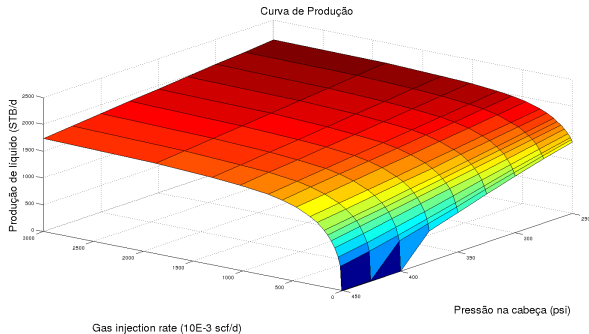
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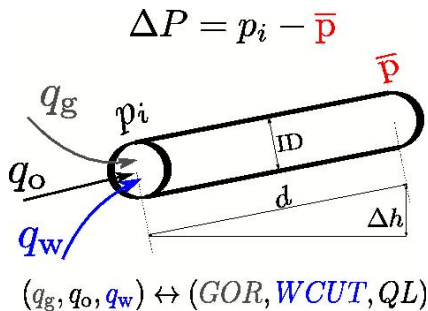
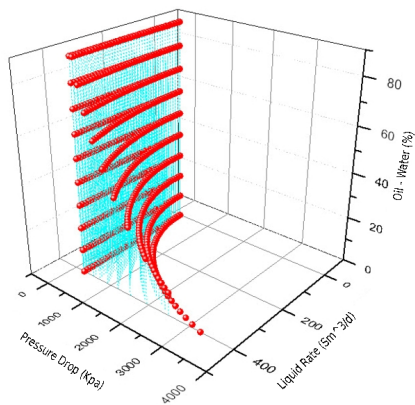
DCC Model

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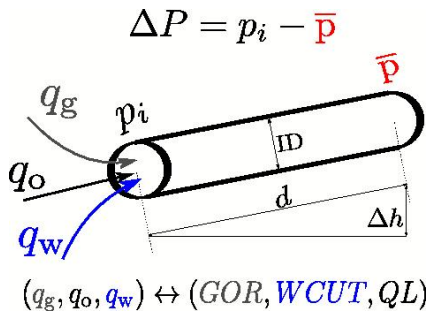
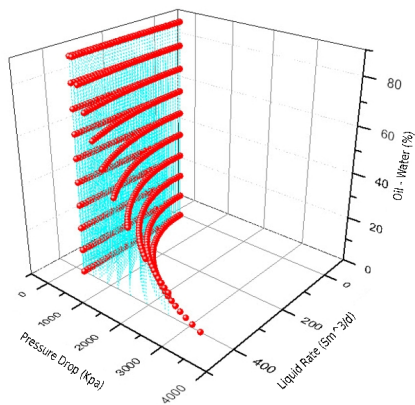
# Well Production $\times$ (Lift-gas, Well-head Pressure).



## Pressure Loss in Flowlines



## Pressure Loss in Flowlines



## Multidimensional Models

Let us consider functions with multidimensional domains:  $f : \mathcal{D} \rightarrow \mathbb{R}$ , in which  $\mathcal{D} \subset \mathbb{R}^n$  is the domain.

An issue that comes up is how to partition the domain.

- ▶ Let  $\mathcal{P}$  be a set of polyhedrons defining a partitioning of  $\mathcal{D}$ :

$$\begin{aligned} \bigcup_{P \in \mathcal{P}} P &= \mathcal{D}, \\ \text{int}(P) \cap P' &= \emptyset, \forall P, P' \in \mathcal{P} \end{aligned}$$

- ▶ The piecewise-linear models can be applied depending on the structure of the partitioning  $\mathcal{P}$  (geometry/topology).

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## Models and Partitioning

These models can be applied to any convex partitioning of the domain:

- ▶ CC, DCC, DLog e Multiple Choice.

Can be applied only to a J-1 triangulation of the domain:

- ▶ Log.

Can be applied only to regular lattices:

- ▶ SOS2.

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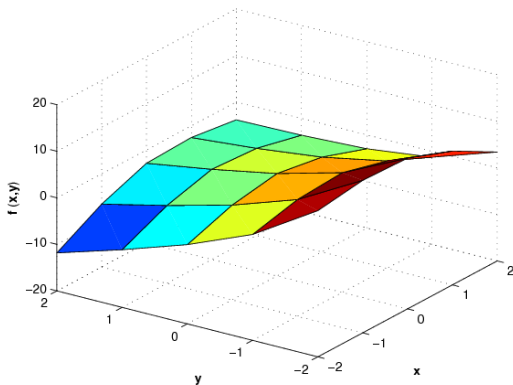
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## Models and Partitioning



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## CC Model

### Definitions:

- ▶  $\mathcal{P}$  is the set of polyhedra defining the domain.
- ▶  $\mathcal{V}$  is the set of breakpoints.
- ▶  $V(P) \subset \mathcal{V}$  is the set of vertices of polyhedron  $P \in \mathcal{P}$ .

### Variables:

- ▶  $\lambda_v \in [0, 1]$  is a weighting variable associated with vertex  $v \in \mathcal{V}$ .
- ▶  $y_P \in \{0, 1\}$  assumes value 1 if  $x \in P$ ,  $P \in \mathcal{P}$ , otherwise assumes value 0.

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## CC Model

$$f = \sum_{v \in \mathcal{V}} f(v) \lambda_v,$$

$$1 = \sum_{v \in \mathcal{V}} \lambda_v,$$

$$\lambda_v \leq \sum_{P \in \mathcal{P}(v)} y_P,$$

$$\lambda_v \geq 0, v \in \mathcal{V},$$

$$y_P \in \{0, 1\}, P \in \mathcal{P}$$

$$x = \sum_{v \in \mathcal{V}} v \lambda_v,$$

$$1 = \sum_{P \in \mathcal{P}} y_P,$$

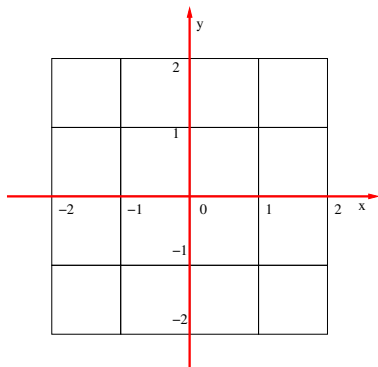
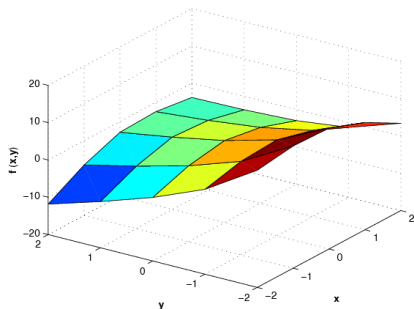
in which:  $\mathcal{P}(v) = \{P \in \mathcal{P} : v \in V(P)\}$  is the set of polyhedra that contain breakpoint  $v$ .



## Sample Function

Consider the nonlinear function:

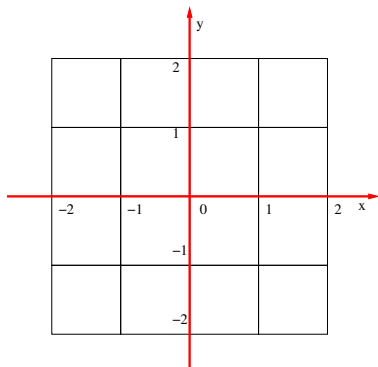
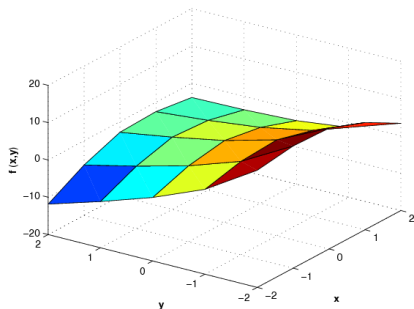
$$f(x, y) = e^{-y} + e^{-x^2} - x^2 + xy - 2y$$



## Sample Function

Consider the nonlinear function:

$$f(x, y) = e^{-y} + e^{-x^2} - x^2 + xy - 2y$$



## Example: CC Model

Vertex set:

$$\mathcal{V} = \{-2, -1, 0, 1, 2\} \times \{-2, -1, 0, 1, 2\}$$

Polyhedra set:

$$\mathcal{P} = \{-2, -1, 0, 1\} \times \{-2, -1, 0, 1\}$$

in which, the index  $(i, j) \in \mathcal{P}$  corresponds to the polyhedron

$$P_{i,j} = \{(x, y) \in \mathbb{R}^2 : i \leq x \leq i+1, j \leq y \leq j+1\}$$

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$$P_{i,j} = \{(x, y) \in \mathbb{R}^2 : i \leq x \leq i+1, j \leq y \leq j+1\}$$

## Example: CC Model

$$\begin{aligned} f &= \sum_{v \in \mathcal{V}} f(v) \lambda_v \\ &= f(-2, -2) \lambda_{(-2, -2)} + f(-2, -1) \lambda_{(-2, -1)} + f(-2, 0) \lambda_{(-2, 0)} \\ &\quad + f(-2, 1) \lambda_{(-2, 1)} + f(-2, 2) \lambda_{(-2, 2)} \\ &\quad \vdots \\ &\quad + f(2, -2) \lambda_{(2, -2)} + f(2, -1) \lambda_{(2, -1)} + f(2, 0) \lambda_{(2, 0)} \\ &\quad + f(2, 1) \lambda_{(2, 1)} + f(2, 2) \lambda_{(2, 2)} \end{aligned}$$

## Example: CC Model

$$\begin{aligned}x &= \sum_{v \in \mathcal{V}} v_x \lambda_v \\ &= (-2)\lambda_{(-2,-2)} + (-2)\lambda_{(-2,-1)} + (-2)\lambda_{(-2,0)} + (-2)\lambda_{(-2,1)} + (-2)\lambda_{(-2,2)} \\ &\quad \vdots \\ &\quad + (2)\lambda_{(2,-2)} + (2)\lambda_{(2,-1)} + (2)\lambda_{(2,0)} + (2)\lambda_{(2,1)} + (2)\lambda_{(2,2)}\end{aligned}$$

## Example: CC Model

$$\begin{aligned} y &= \sum_{v \in \mathcal{V}} v_y \lambda_v \\ &= (-2)\lambda_{(-2,-2)} + (-1)\lambda_{(-2,-1)} + (0)\lambda_{(-2,0)} + (1)\lambda_{(-2,1)} + (2)\lambda_{(-2,2)} \\ &\quad \vdots \\ &\quad + (-2)\lambda_{(2,-2)} + (-1)\lambda_{(2,-1)} + (0)\lambda_{(2,0)} + (1)\lambda_{(2,1)} + (2)\lambda_{(2,2)} \end{aligned}$$

## Example: CC Model

$$\begin{aligned} 1 &= \sum_{v \in \mathcal{V}} \lambda_v \\ &= \lambda_{(-2,-2)} + \lambda_{(-2,-1)} + \lambda_{(-2,0)} + \lambda_{(-2,1)} + \lambda_{(-2,2)} \\ &\quad + \lambda_{(-1,-2)} + \lambda_{(-1,-1)} + \lambda_{(-1,0)} + \lambda_{(-1,1)} + \lambda_{(-1,2)} \\ &\quad \vdots \\ &\quad + \lambda_{(2,-2)} + \lambda_{(2,-1)} + \lambda_{(2,0)} + \lambda_{(2,1)} + \lambda_{(2,2)} \end{aligned}$$



## Example: CC Model

$$\begin{aligned} 1 &= \sum_{P \in \mathcal{P}} y_P \\ &= y_{(-2,-2)} + y_{(-2,-1)} + y_{(-2,0)} + y_{(-2,1)} \\ &\quad + y_{(-1,-2)} + y_{(-1,-1)} + y_{(-1,0)} + y_{(-1,1)} \\ &\quad \vdots \\ &\quad + y_{(2,-2)} + y_{(2,-1)} + y_{(2,0)} + y_{(2,1)} \end{aligned}$$

## Example: CC Model

The constraint family

$$\lambda_v \leq \sum_{P \in \mathcal{P}(v)} y_P,$$

becomes

$$\lambda_{(-2,-2)} \leq y_{(-2,-2)}$$

$$\lambda_{(-2,-1)} \leq y_{(-2,-2)} + y_{(-2,-1)}$$

$$\lambda_{(-2,0)} \leq y_{(-2,-1)} + y_{(-2,0)}$$

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$$\vdots \quad \quad \quad \vdots$$

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$$\lambda_{(-1,2)} \leq y_{(-2,1)} + y_{(-1,1)}$$

$$\vdots \quad \quad \quad \vdots$$

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**DCC Model**

SOS2 Model

## Disaggregated Convex Combination (DCC) Model

$$f = \sum_{P \in \mathcal{P}} \sum_{v \in \mathcal{V}(P)} f(v) \lambda_{P,v},$$

$$x = \sum_{P \in \mathcal{P}} \sum_{v \in \mathcal{V}(P)} v \lambda_{P,v},$$

$$y_P = \sum_{v \in \mathcal{V}(P)} \lambda_{P,v}, \quad P \in \mathcal{P},$$

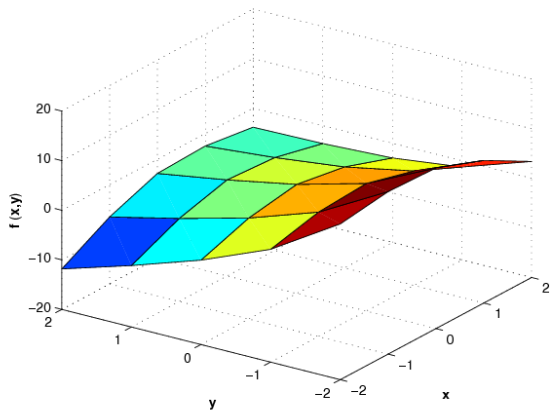
$$\mathbf{1} = \sum_{P \in \mathcal{P}} y_P,$$

$$\lambda_{P,v} \geq 0, \quad P \in \mathcal{P}, \quad v \in \mathcal{V}(P),$$

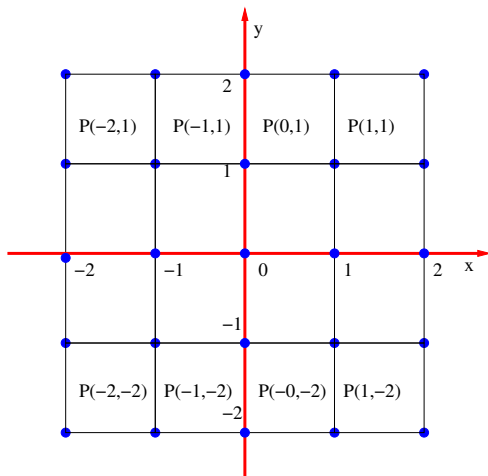
$$y_P \in \{0, 1\}, \quad P \in \mathcal{P}$$

where:  $\lambda_{P,v}$  is associated to vertex  $v$  of polyhedron  $P$  (breakpoint).

## Example: Function



## Example: Domain



## Example: DCC Model

$$\begin{aligned} f &= \sum_{P \in \mathcal{P}} \sum_{v \in \mathcal{V}(P)} f(v) \lambda_{P,v}, \\ &= f(-2, -2) \lambda_{(-2,-2)}^{P(-2,-2)} + f(-2, -1) \lambda_{(-2,-1)}^{P(-2,-2)} \\ &\quad + f(-1, -2) \lambda_{(-1,-2)}^{P(-2,-2)} + f(-1, -1) \lambda_{(-1,-1)}^{P(-2,-2)} \\ &\quad + f(-1, -2) \lambda_{(-1,-2)}^{P(-1,-2)} + f(-1, -1) \lambda_{(-1,-1)}^{P(-1,-2)} \\ &\quad + f(0, -2) \lambda_{(0,-2)}^{P(-1,-2)} + f(0, -1) \lambda_{(0,-1)}^{P(-1,-2)} \\ &\quad \vdots \\ &\quad + f(1, 1) \lambda_{(1,1)}^{P(1,1)} + f(1, 2) \lambda_{(1,2)}^{P(1,1)} \\ &\quad + f(2, 1) \lambda_{(2,1)}^{P(1,1)} + f(2, 2) \lambda_{(2,2)}^{P(1,1)} \end{aligned}$$

## Example: DCC Model

$$\begin{aligned}(x, y) &= \sum_{P \in \mathcal{P}} \sum_{v \in \mathcal{V}(P)} v \lambda_{P,v}, \\ &= (-2, -2) \lambda_{(-2,-2)}^{P(-2,-2)} + (-2, -1) \lambda_{(-2,-1)}^{P(-2,-2)} \\ &\quad + (-1, -2) \lambda_{(-1,-2)}^{P(-2,-2)} + (-1, -1) \lambda_{(-1,-1)}^{P(-2,-2)} \\ &\quad + (-1, -2) \lambda_{(-1,-2)}^{P(-1,-2)} + (-1, -1) \lambda_{(-1,-1)}^{P(-1,-2)} \\ &\quad + (0, -2) \lambda_{(0,-2)}^{P(-1,-2)} + (0, -1) \lambda_{(0,-1)}^{P(-1,-2)} \\ &\quad \vdots \\ &\quad + (1, 1) \lambda_{(1,1)}^{P(1,1)} + (1, 2) \lambda_{(1,2)}^{P(1,1)} \\ &\quad + (2, 1) \lambda_{(2,1)}^{P(1,1)} + (2, 2) \lambda_{(2,2)}^{P(1,1)}\end{aligned}$$



## Example: DCC Model

The constraint family

$$y_P = \sum_{v \in \mathcal{V}(P)} \lambda_{P,v}, \quad P \in \mathcal{P},$$

becomes

$$\begin{aligned} y_{P(-2,-2)} &= \lambda_{(-2,-2)}^{P(-2,-2)} + \lambda_{(-2,-1)}^{P(-2,-2)} + \lambda_{(-1,-2)}^{P(-2,-2)} + \lambda_{(-1,-1)}^{P(-2,-2)} \\ y_{P(-1,-2)} &= \lambda_{(-1,-2)}^{P(-1,-2)} + \lambda_{(-1,-1)}^{P(-1,-2)} + \lambda_{(0,-2)}^{P(-1,-2)} + \lambda_{(0,-1)}^{P(-1,-2)} \\ &\vdots \quad \quad \quad \vdots \\ y_{P(1,1)} &= \lambda_{(1,1)}^{P(1,1)} + \lambda_{(1,2)}^{P(1,1)} + \lambda_{(2,1)}^{P(1,1)} + \lambda_{(2,2)}^{P(1,1)} \end{aligned}$$

## Example: DCC Model

The constraint family

$$1 = \sum_{P \in \mathcal{P}} y_P,$$

becomes

$$1 = y_{P(-2,-2)} + y_{P(-1,-2)} + y_{P(0,-2)} + \dots + y_{P(1,1)}$$

# Summary

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# SOS2 Model

## Objectives:

- ▶ Piecewise-linear approximation of the function  $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ .

## Definitions:

- ▶ Breakpoint set in  $x : \mathcal{X} = \{x_1, x_2, \dots, x_n\}$ .
- ▶ Breakpoint set in  $y : \mathcal{Y} = \{y_1, y_2, \dots, y_n\}$ .
- ▶ Given the values of the function  $f$  for  $(x, y) \in \mathcal{X} \times \mathcal{Y}$ .
- ▶ Polyhedron set given by:

$$\mathcal{P} = \{[x_i, x_{i+1}] \times [y_j, y_{j+1}] : i = 1, \dots, n-1, j = 1, \dots, n-1\}$$

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## SOS2 Model

$$f = \sum_{x \in \mathcal{X}} \sum_{y \in \mathcal{Y}} f(x, y) \lambda_{x, y},$$

$$x = \sum_{x \in \mathcal{X}} \sum_{y \in \mathcal{Y}} x \lambda_{x, y},$$

$$y = \sum_{x \in \mathcal{X}} \sum_{y \in \mathcal{Y}} y \lambda_{x, y},$$

$$1 = \sum_{x \in \mathcal{X}} \sum_{y \in \mathcal{Y}} \lambda_{x, y},$$

$$\eta_x = \sum_{y \in \mathcal{Y}} \lambda_{x, y}, \quad \forall x \in \mathcal{X},$$

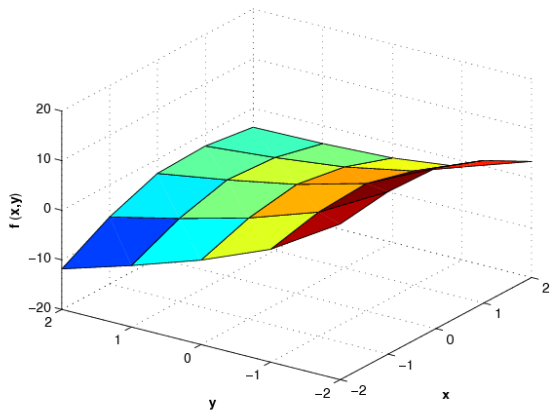
$(\eta_x)_{x \in \mathcal{X}}$  is SOS2

$$\lambda_{x, y} \geq 0, \quad \forall x \in \mathcal{X}, y \in \mathcal{Y}$$

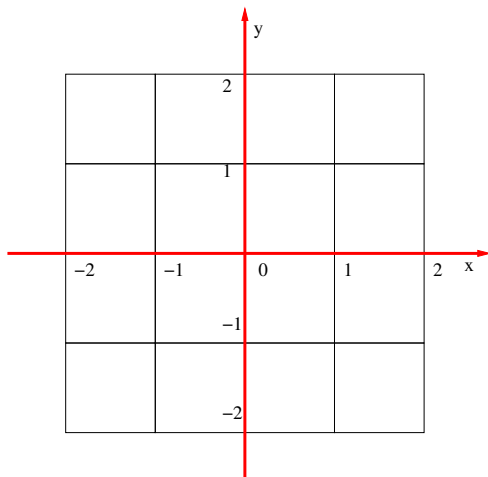
$$\eta_y = \sum_{x \in \mathcal{X}} \lambda_{x, y}, \quad \forall y \in \mathcal{Y},$$

$(\eta_y)_{y \in \mathcal{Y}}$  is SOS2

## Example: Function



## Example: Domain





## Example: SOS2 Model

Function approximation becomes

$$\begin{aligned} f &= \sum_{x \in \mathcal{X}} \sum_{y \in \mathcal{Y}} f(x, y) \lambda_{x,y} \\ &= +f(-2, -2)\lambda_{(-2,-2)} + f(-2, -1)\lambda_{(-2,-1)} + f(-2, 0)\lambda_{(-2,0)} \\ &\quad + f(-2, 1)\lambda_{(-2,1)} + f(-2, 2)\lambda_{(-2,2)} \\ &\quad \vdots \\ &\quad + f(2, -2)\lambda_{(2,-2)} + f(2, -1)\lambda_{(2,-1)} + f(2, 0)\lambda_{(2,0)} \\ &\quad + f(2, 1)\lambda_{(2,1)} + f(2, 2)\lambda_{(2,2)} \end{aligned}$$

## Example: SOS2 Model

Domain variables

$$\begin{aligned}(x, y) &= \sum_{x \in \mathcal{X}} \sum_{y \in \mathcal{Y}} (x, y) \lambda_{x,y} \\ &= +(-2, -2) \lambda_{(-2,-2)} + (-2, -1) \lambda_{(-2,-1)} + (-2, 0) \lambda_{(-2,0)} \\ &\quad + (-2, 1) \lambda_{(-2,1)} + (-2, 2) \lambda_{(-2,2)} \\ &\quad \vdots \\ &\quad + (2, -2) \lambda_{(2,-2)} + (2, -1) \lambda_{(2,-1)} + (2, 0) \lambda_{(2,0)} \\ &\quad + (2, 1) \lambda_{(2,1)} + (2, 2) \lambda_{(2,2)}\end{aligned}$$

## Example: SOS2 Model

$$\begin{aligned} 1 &= \sum_{x \in \mathcal{X}} \sum_{y \in \mathcal{Y}} \lambda_{x,y} \\ &= \lambda_{(-2,-2)} + \lambda_{(-2,-1)} + \lambda_{(-2,0)} + \lambda_{(-2,1)} + \lambda_{(-2,2)} \\ &\quad \vdots \\ &\quad + \lambda_{(2,-2)} + \lambda_{(2,-1)} + \lambda_{(2,0)} + \lambda_{(2,1)} + \lambda_{(2,2)} \end{aligned}$$

# SOS2 Model

The constraint family

$$\eta_x = \sum_{y \in \mathcal{Y}} \lambda_{x,y}, \quad x \in \mathcal{X}$$

becomes

$$\eta_{-2}^x = \lambda_{(-2,-2)} + \lambda_{(-2,-1)} + \lambda_{(-2,0)} + \lambda_{(-2,1)} + \lambda_{(-2,2)}$$

$$\eta_{-1}^x = \lambda_{(-1,-2)} + \lambda_{(-1,-1)} + \lambda_{(-1,0)} + \lambda_{(-1,1)} + \lambda_{(-1,2)}$$

$$\vdots = \quad \quad \quad \vdots$$

$$\eta_2^x = \lambda_{(2,-2)} + \lambda_{(2,-1)} + \lambda_{(2,0)} + \lambda_{(2,1)} + \lambda_{(2,2)}$$

# SOS2 Model

The constraint family

$$\eta_y = \sum_{x \in \mathcal{X}} \lambda_{x,y}, \quad y \in \mathcal{Y}$$

becomes

$$\eta_{-2}^y = \lambda_{(-2,-2)} + \lambda_{(-1,-2)} + \lambda_{(0,-2)} + \lambda_{(1,-2)} + \lambda_{(2,-2)}$$

$$\eta_{-1}^y = \lambda_{(-2,-1)} + \lambda_{(-1,-1)} + \lambda_{(0,-1)} + \lambda_{(1,-1)} + \lambda_{(2,-1)}$$

$$\vdots = \quad \quad \quad \vdots$$

$$\eta_2^y = \lambda_{(-2,2)} + \lambda_{(-1,2)} + \lambda_{(0,2)} + \lambda_{(1,2)} + \lambda_{(2,2)}$$

## Piecewise-Linear Approximation: Multidimensional

- ▶ End!
- ▶ Thank you for your attention.