

# Optimality of Relaxations for Integer-Constrained Problems

Jan Lellmann

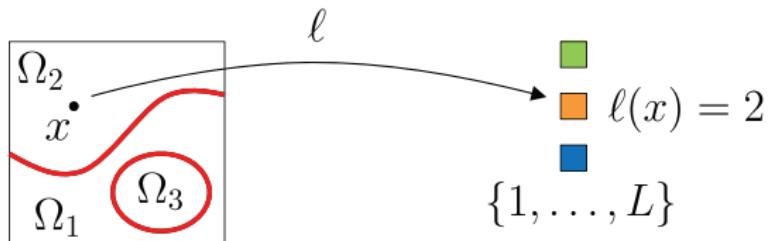
DAMTP, University of Cambridge

Joint work with: F. Lenzen, C. Schnörr (IPA/HCI, University of Heidelberg)

SIAM IS12, May 2012

# Motivation – Problem

- ▶ Labeling problem:

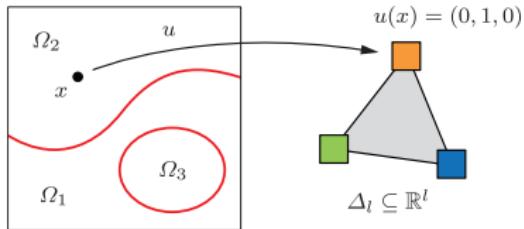


- ▶ Partition image domain  $\Omega$  into  $L$  regions
- ▶ Discrete decision at each point in *continuous* domain  $\Omega$
- ▶ Variational Approach:

$$\min_{\ell} \underbrace{\int_{\Omega} s(\ell(x), x) dx}_{\text{local data fidelity}} + \underbrace{J(\ell)}_{\text{regularizer}}$$

# Model – Multi-Class Labeling

- Multi-class relaxation: [Lie et al. 06, Zach et al. 08, Lellmann et al. 09, Pock et al. 09]



- Embed labels into  $\mathbb{R}^L$  as  $\mathcal{E} := \{e^1, \dots, e^L\}$ , relax integrality constraint to the unit simplex:

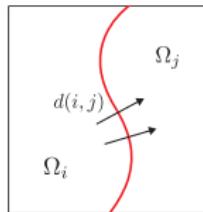
$$\Delta_L := \{x \in \mathbb{R}^L \mid x \geq 0, \sum_i x_i = 1\} = \text{conv } \mathcal{E},$$

$$\min_{u \in \text{BV}(\Omega, \Delta_L)} f(u), \quad f(u) := \int_{\Omega} \langle u(x), s(x) \rangle dx + \int_{\Omega} \Psi(Du)$$

- Advantages: No explicit parametrization, rotation invariance, convex

# Model – Envelope Relaxation

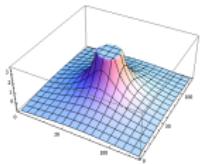
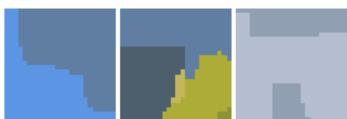
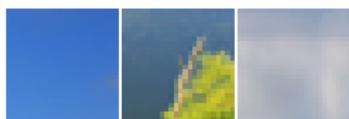
- ▶  $J(\ell)$ : Weight boundary length by *interaction potential*  $d(i,j)$



- ▶  $\psi$  implicitly defined for given  $d$  [ChambolleCremersPock08,LellmannSchnoerr10]  $\rightarrow$  *metric labeling*.

# Motivation – BV formulation

- ▶ Spatially continuous formulation avoids metrification artifacts:



# Model – Rounding

- ▶ *Fractional* solutions may occur:



- ▶ **Goal:** Find *rounding scheme*  $u^* \mapsto \bar{u}^* : \Omega \rightarrow \{e^1, \dots, e^L\}$  such that

$$f(\underbrace{\bar{u}^*}_{\text{rounded relaxed solution}}) \leq C f(\underbrace{u_{\mathcal{E}}^*}_{\text{best integral solution}}).$$

for some  $C \geq 1$ .

# Rounding – Generalized Coarea Formula

- ▶ Two-class case: Generalized *coarea formula* [Strang83, ChanEsedogluNikolova06, Zach et al. 09, Olsson et al. 09]

$$f(u) = \int_0^1 f(\bar{u}_\gamma) d\gamma, \quad \bar{u}_\gamma := \begin{cases} e^1, & u_1(x) > \gamma, \\ e^2, & u_1(x) \leq \gamma. \end{cases}$$

- ▶ Also: *Choquet integral, Lovász extension, levelable function,...*

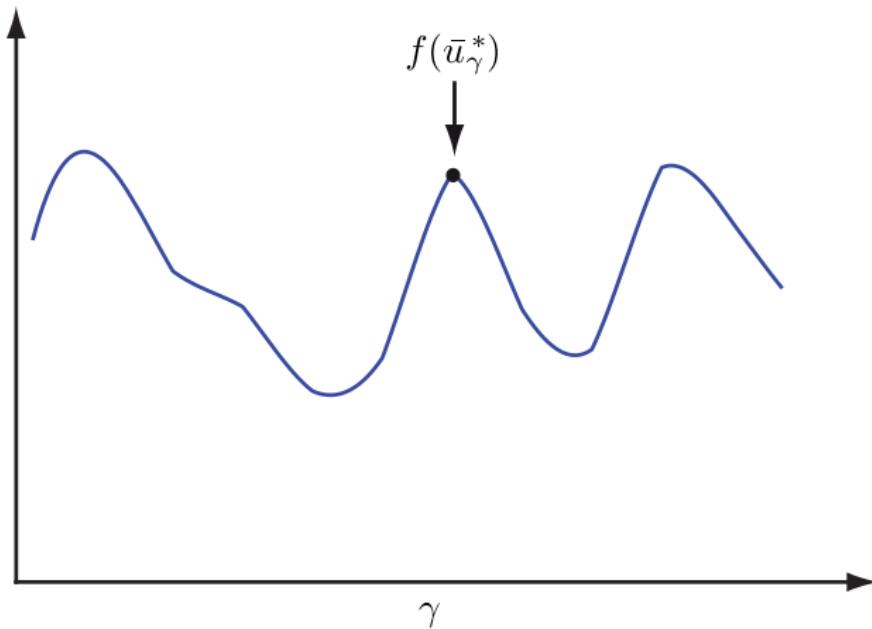
# Rounding – Multi-Class Case

- ▶ *Probabilistic interpretation:*

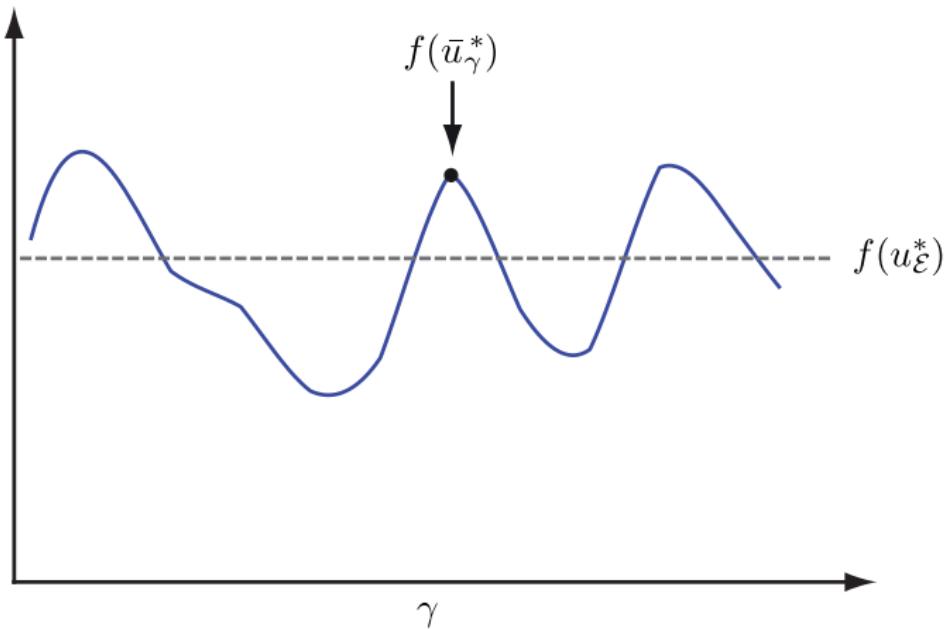
$$f(u) = \int_0^1 f(\bar{u}_\gamma) d\gamma = \mathbb{E}_\gamma f(\bar{u}_\gamma).$$

- ▶ Rounding step  $u \mapsto \bar{u}_\gamma$  does not increase  $f$  *in the expectation*
- ▶ Consequence: global *integral* minimizer for a.e.  $\gamma!$  [ChanEsedogluNikolova06]

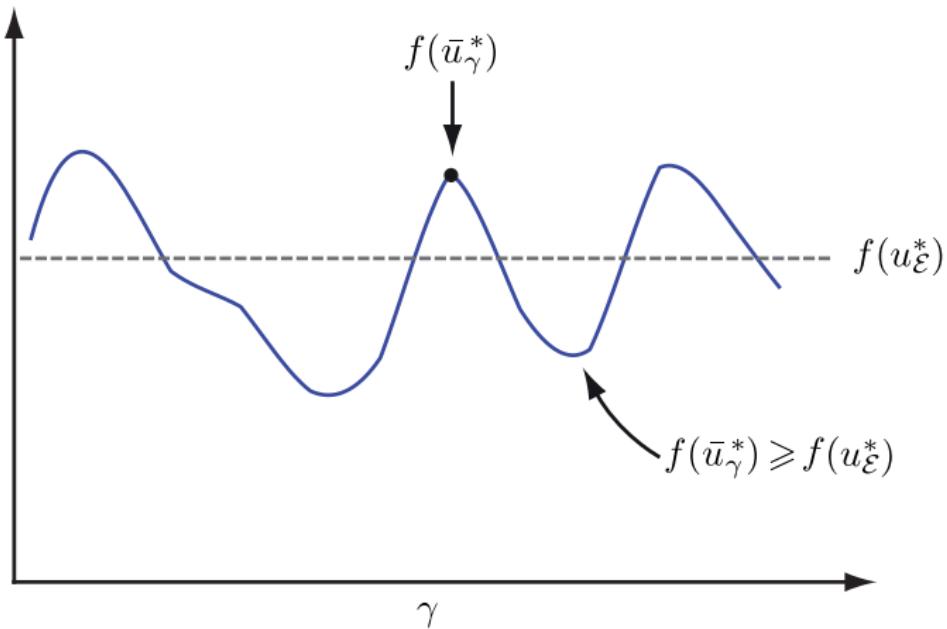
# Rounding – Generalized Coarea Formula



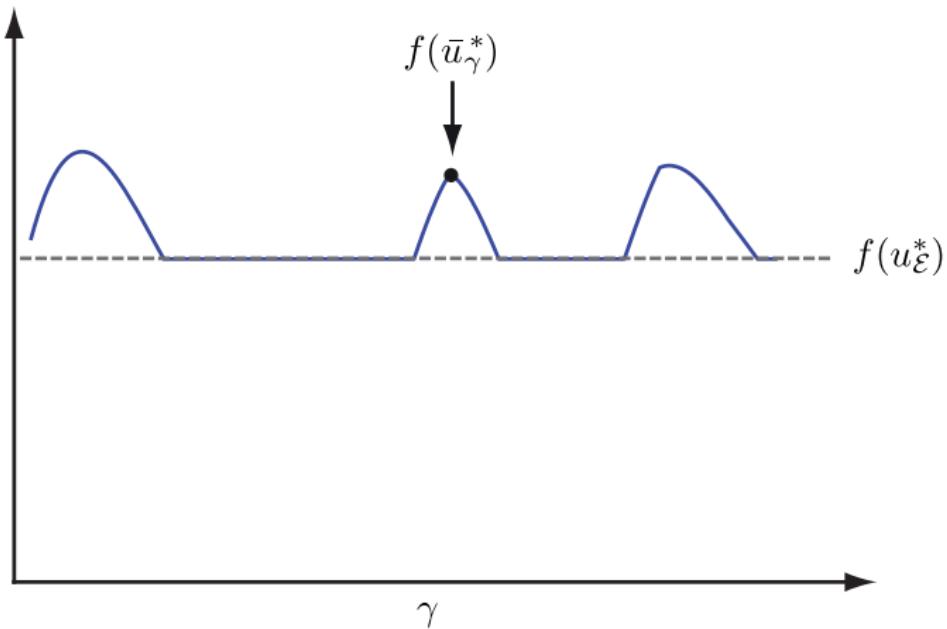
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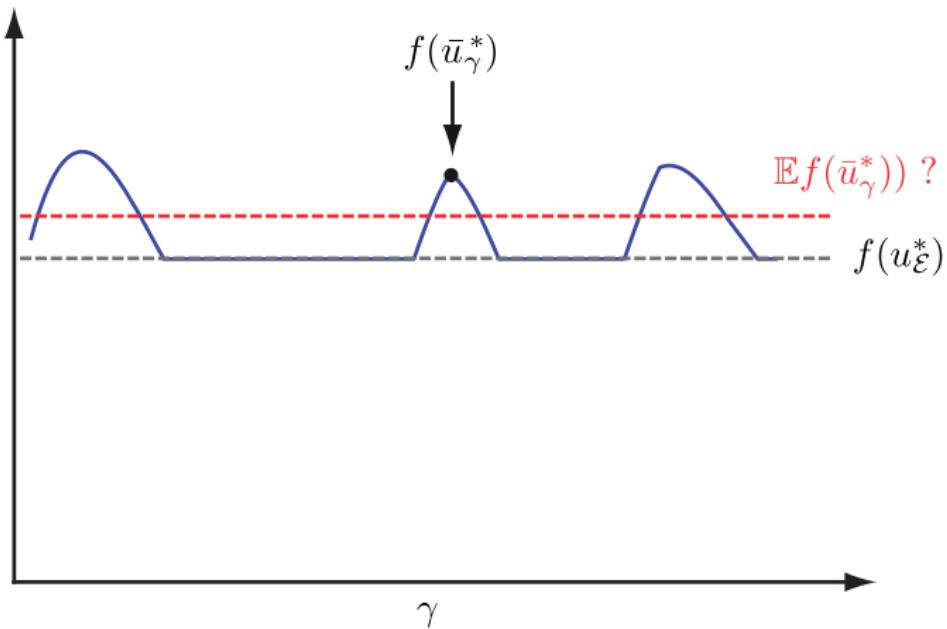
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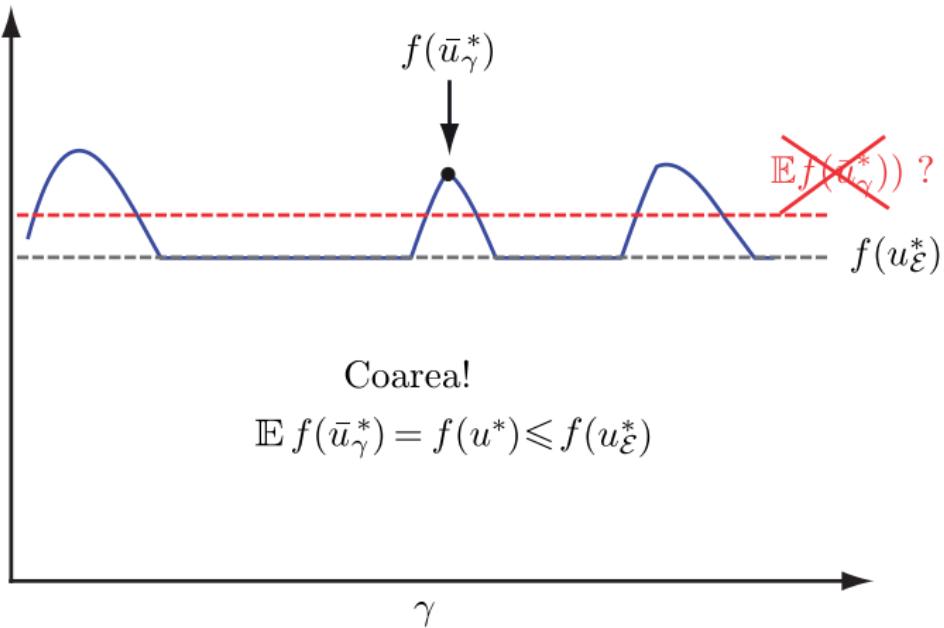
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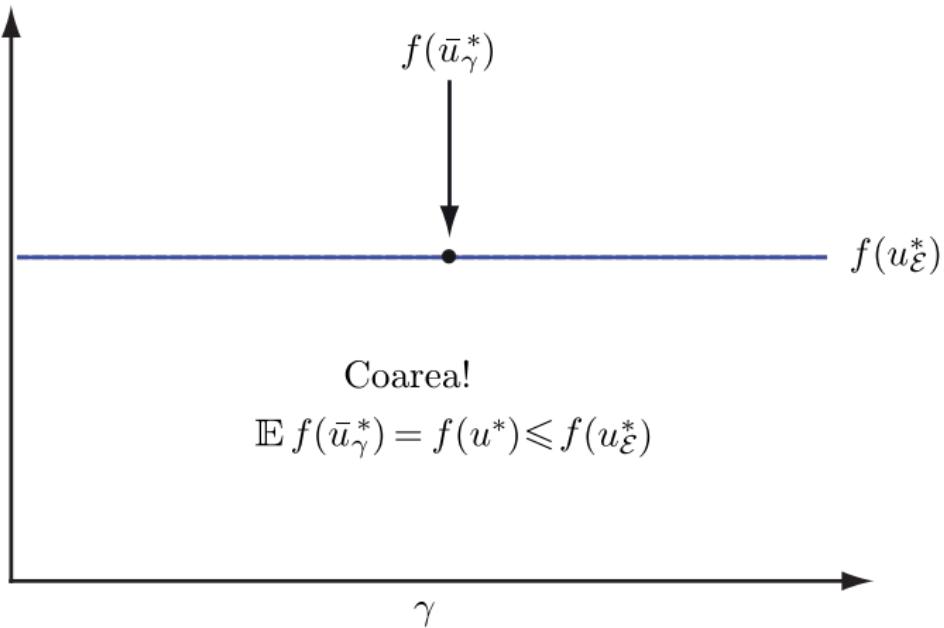
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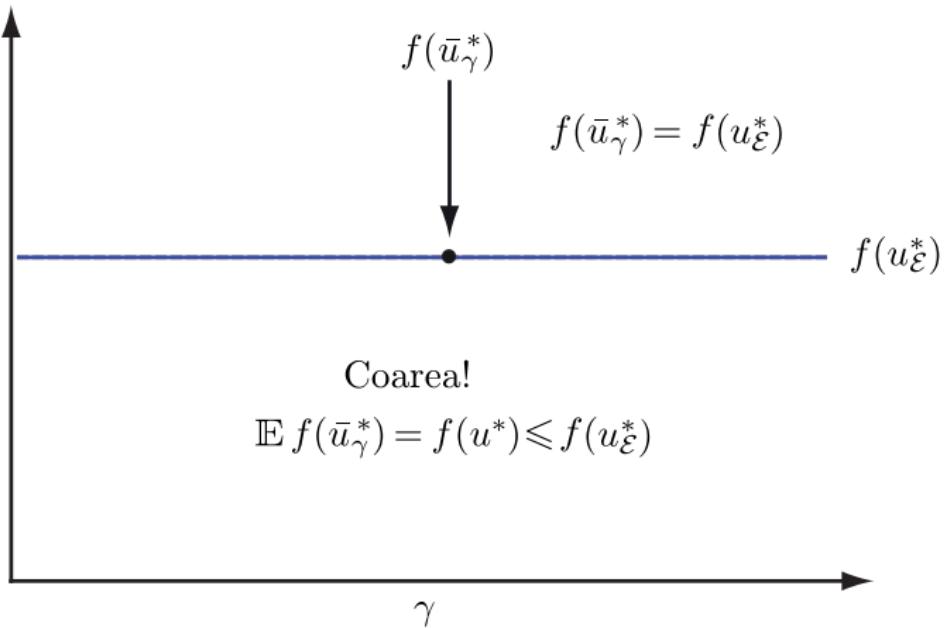
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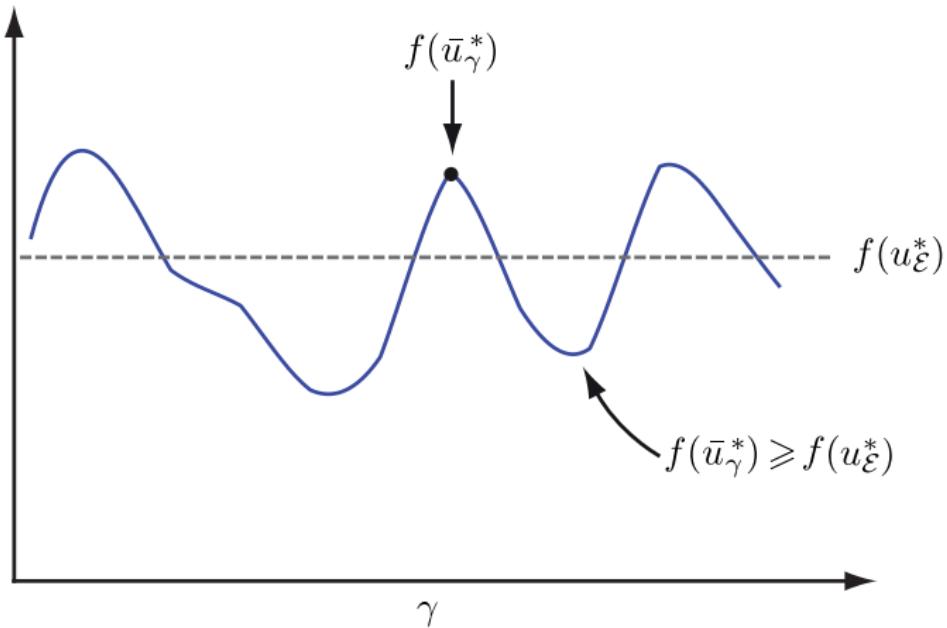
# Rounding – Approximate Coarea Formula

- ▶ Multi-class generalization (*approximate generalized coarea formula*):

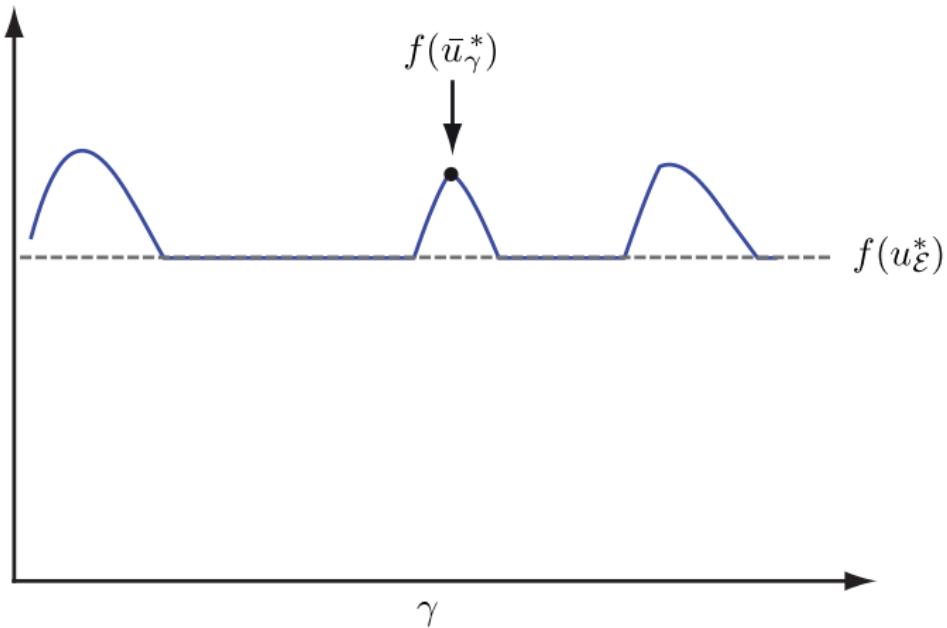
$$\textcolor{blue}{C}f(u) \geq \int_{\Gamma} f(\bar{u}_{\gamma}) d\mu(\gamma) = \mathbb{E}_{\gamma} f(\bar{u}_{\gamma})$$

- ▶ Rounding step  $u \mapsto \bar{u}_{\gamma}$  does not increase  $f$  too much *in the expectation*
- ▶ No bounds for *individual*  $\gamma$ !

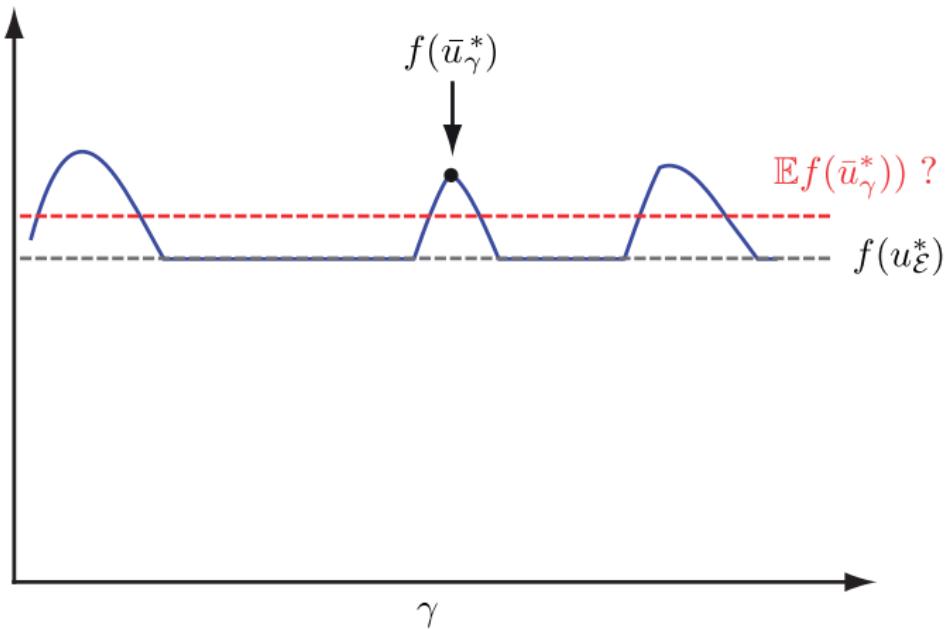
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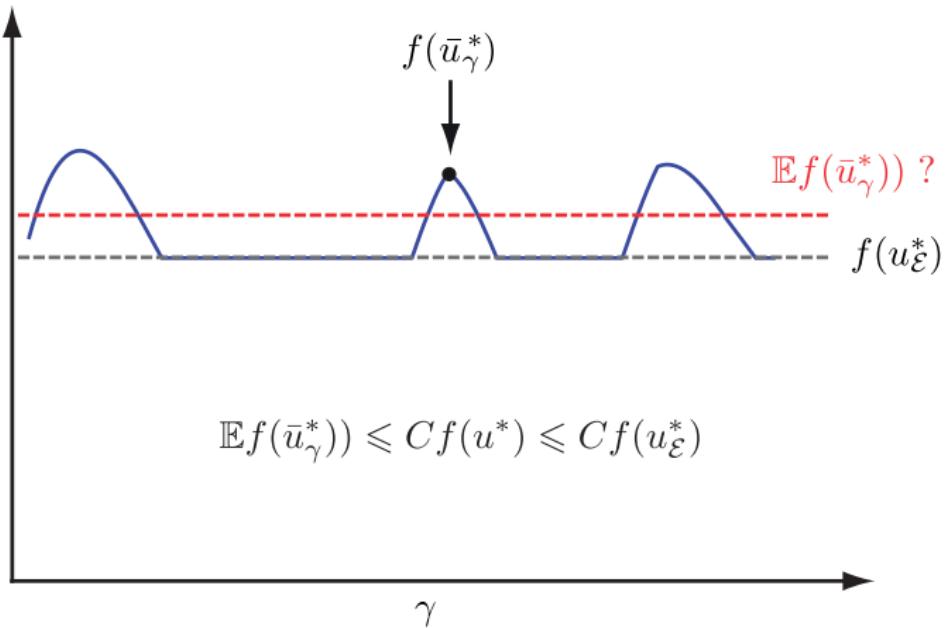
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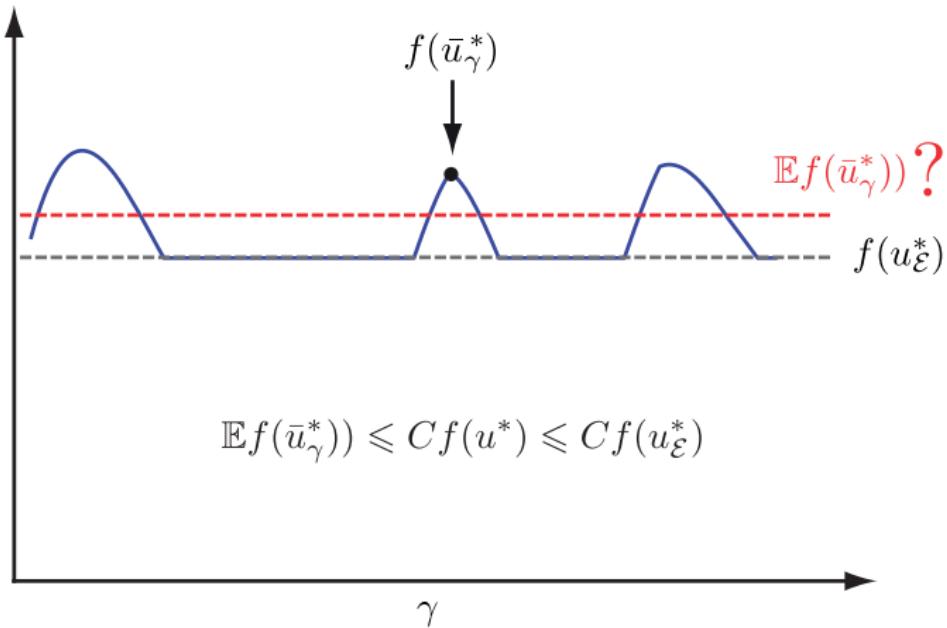
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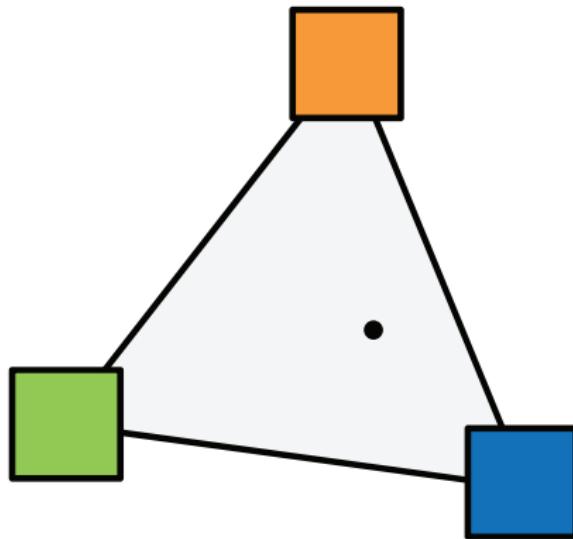
# Rounding – Multi-Class Case

- ▶ Multi-class generalization (*approximate generalized coarea formula*):

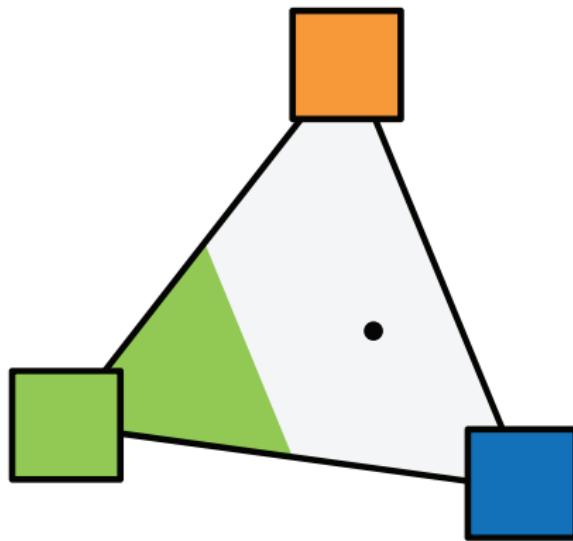
$$Cf(u) \geq \int_{\Gamma} f(\bar{u}_\gamma) d\mu(\gamma) = \mathbb{E}_\gamma f(\bar{u}_\gamma)$$

- ▶ Need to define:
  - ▶ *parameter space*  $\Gamma$
  - ▶ parametrized *rounding method*  $u \mapsto \bar{u}_\gamma$
  - ▶ *probability measure*  $\mu$  on  $\Gamma$
  - ▶ bound  $C$  independent of input

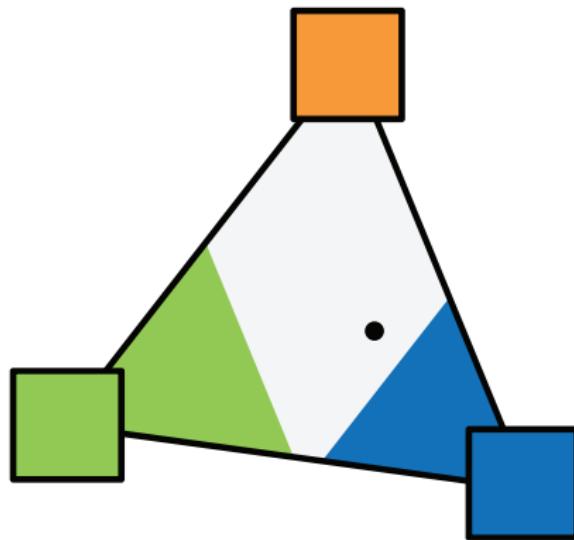
# Optimality – Example



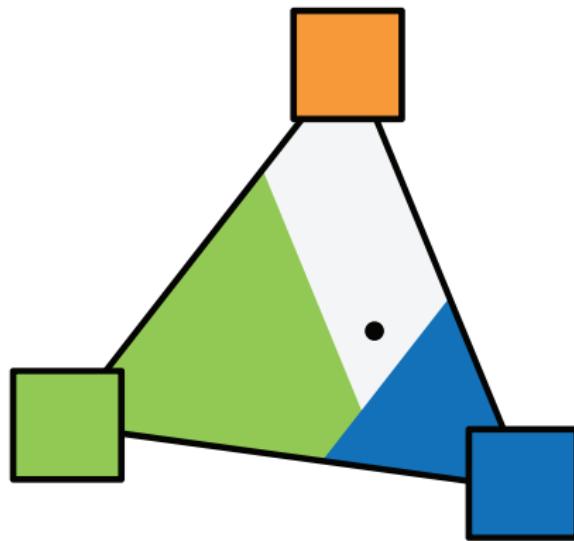
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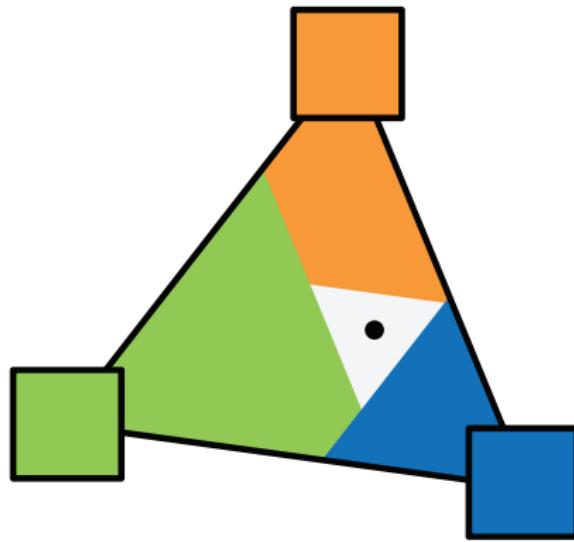
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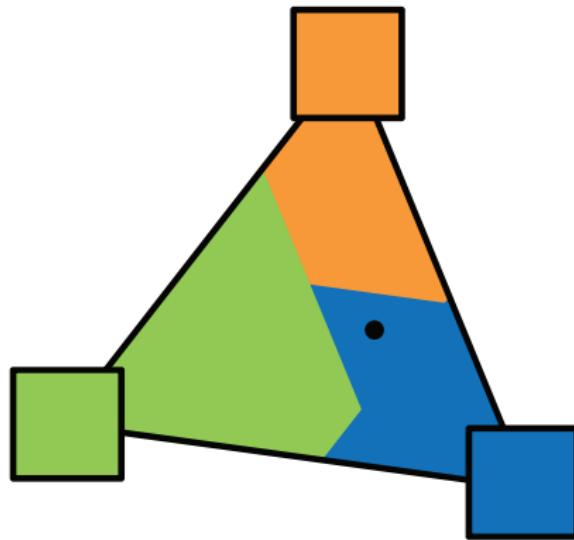
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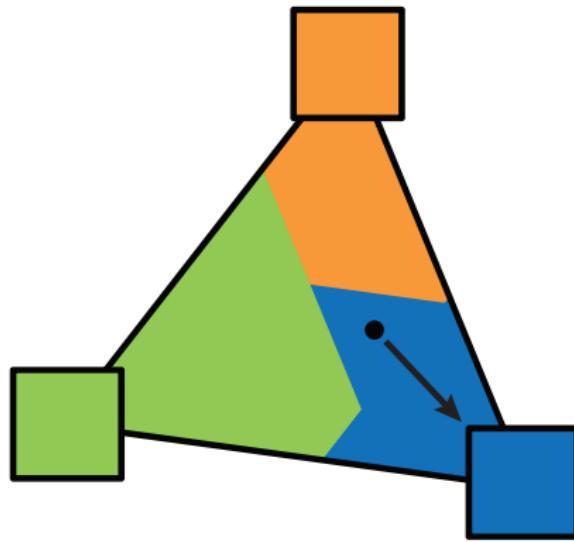
# Optimality – Example



# Optimality – Example



# Optimality – Example



# Optimality – Randomized Rounding

## Algorithm 1 (Randomized Rounding in BV)

1. **Input:**  $u^0 \in \text{BV}(\Omega, \Delta_L)$
2. **For**  $k = 1, 2, \dots$
3.     Sample  $\gamma^k = (i^k, \alpha^k) \in \{1, \dots, L\} \times [0, 1]$  uniformly
4.      $u^k \leftarrow \begin{cases} e^{i^k}, & u_{i^k}^{k-1} > \alpha^k, \\ u^{k-1}, & u_{i^k}^{k-1} \leq \alpha^k. \end{cases}$
5. **Output:** Limit  $\bar{u}$  of  $(u^k)$

► Parameter space: *sequences*  $\gamma \in \Gamma := (\{1, \dots, L\} \times [0, 1])^{\mathbb{N}}$

# Optimality – Termination

## Theorem (Termination)

*Let  $u \in \text{BV}(\Omega, \Delta_L)$ . Then (almost surely) Alg. 1 generates a sequence that becomes stationary in some  $\bar{u} \in \text{BV}(\Omega, \mathcal{E})$ .*

- ▶ Result is in BV
- ▶ Independent of data term

# Randomized Rounding – Motivation

- ▶ In finite-dimensional setting: Can show for uniform metric

[KleinbergTardos02]:

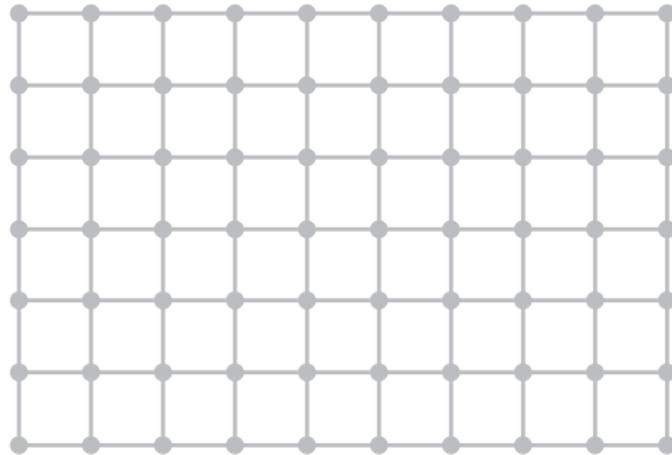
$$\mathbb{E}_\gamma f(\bar{u}_\gamma) \leqslant 2f(u).$$

- ▶ Similar bounds:

- ▶ multiway cut [Dahlhaus et al. 94]
- ▶  $\alpha$ -expansion [Boykov et al. 01]
- ▶ LP relaxation [KomodakisTziritas 07]

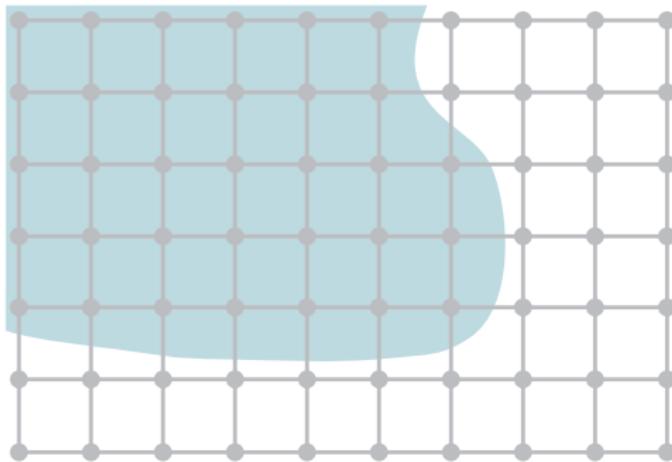
# Optimality – Proof

$$E := \{x \in \Omega \mid u_{ik} > \alpha^k\}$$



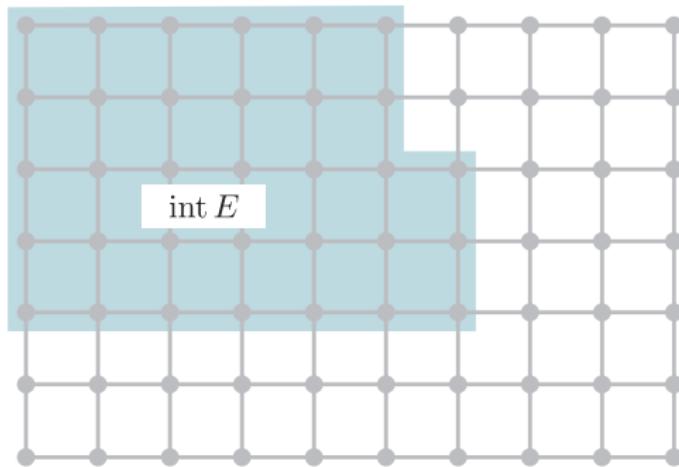
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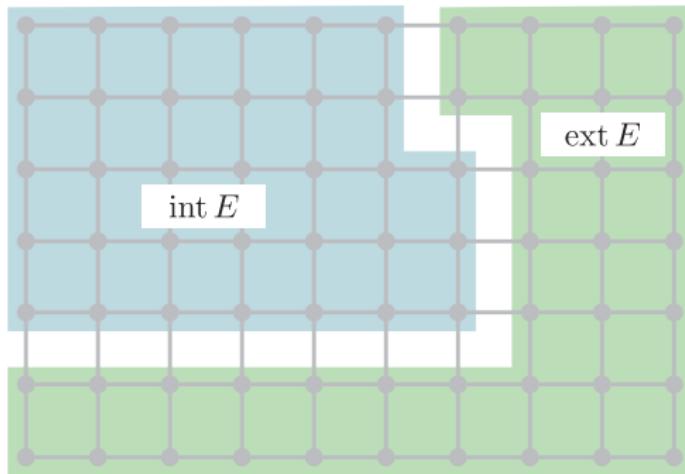
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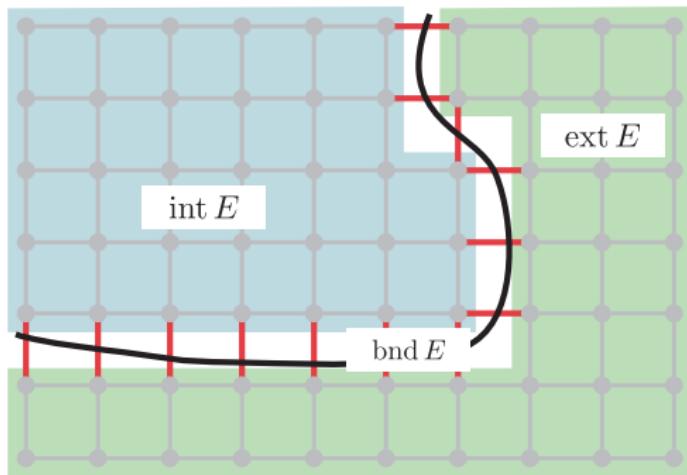
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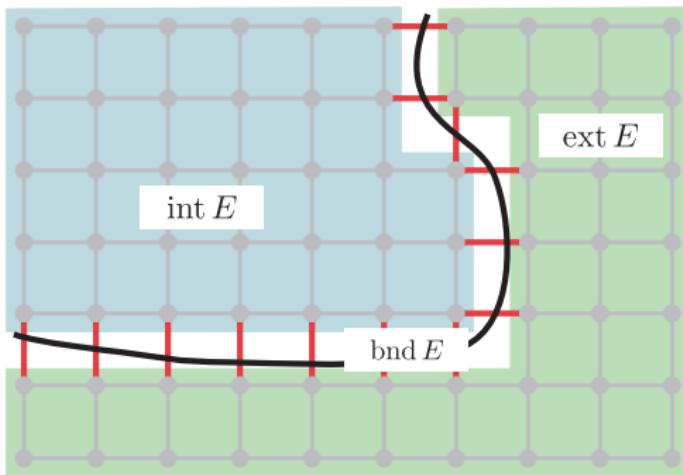
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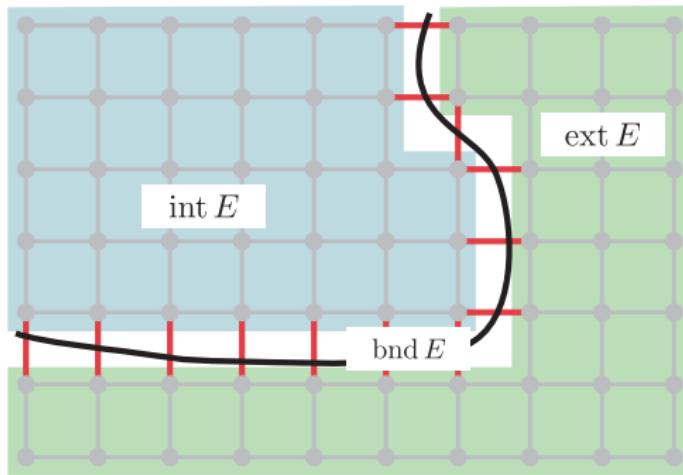
$$E := \{x \in \Omega \mid u_{ik} > \alpha^k\}$$



$$\mathbb{R}^n : \quad J(u) \leq J|_{\text{int } E}(u) + J|_{\text{ext } E}(u) + c \operatorname{Per}(E)$$

# Optimality – Proof

$$E := \{x \in \Omega \mid u_{ik} > \alpha^k\}$$



$$\mathbb{R}^n : J(u) \leq J|_{\text{int } E}(u) + J|_{\text{ext } E}(u) + c \operatorname{Per}(E)$$

$$\text{BV}(\Omega) : |\psi(Du)|(\Omega) \leq |\psi(Du)|(E^1) + |\psi(Du)|(E^0) + c \operatorname{Per}(E),$$

$$(E)^t := \{x \in \Omega \mid \lim_{\rho \searrow 0} \frac{|\mathcal{B}_\rho(x) \cap E|}{|\mathcal{B}_\rho(x)|} = t\}, \quad t \in [0, 1].$$

# Optimality – Main Result

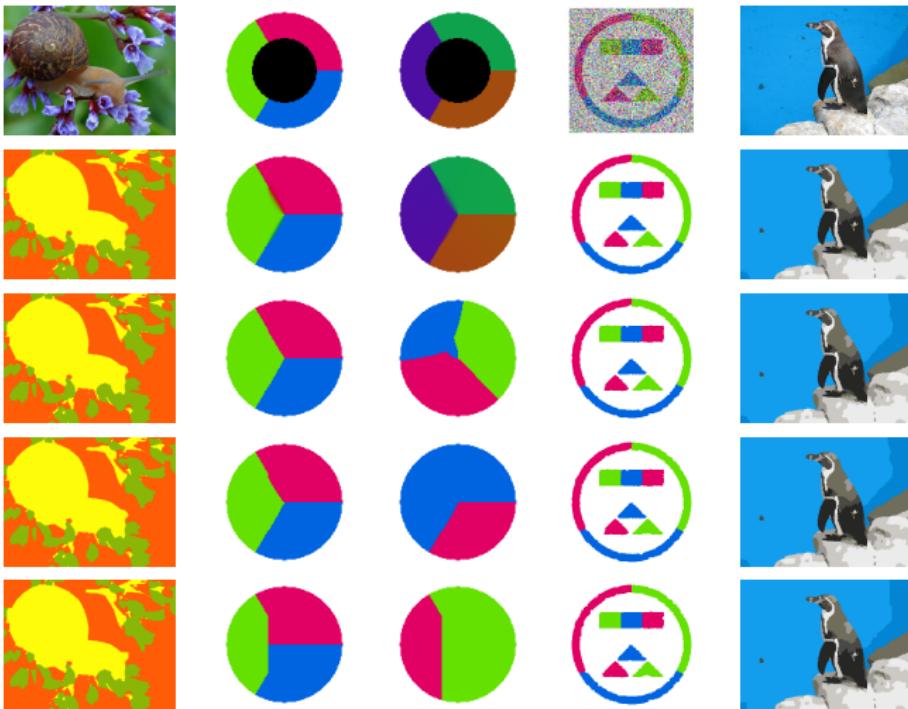
Theorem (Optimality [LellmannLenzenSchnoerr2011])

Let  $u \in \text{BV}(\Omega, \Delta_L)$ ,  $s \in L^\infty(\Omega)^L$ ,  $s \geq 0$ ,  $d$  metric. Then

$$\mathbb{E}f(\bar{u}) \leq 2 \frac{\max_{i \neq j} d(i, j)}{\min_{i \neq j} d(i, j)} f(u) \quad \text{and} \quad \mathbb{E}f(\bar{u}^*) \leq 2 \frac{\max_{i \neq j} d(i, j)}{\min_{i \neq j} d(i, j)} f(u^*_\varepsilon).$$

- ▶ Provides “approximate” generalized coarea formula
- ▶ Compatible with bounds for finite-dimensional multiway cut,  
 $\alpha$ -expansion, LP relaxation
- ▶ Formulated in BV, independent of discretization, true *a priori* bound

# Experiments

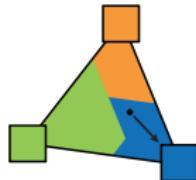


# Experiments – Results

problem	1	2	3	4	5	6	7	8	9	10
# points	76800	14400	14400	129240	76800	86400	86400	76800	86400	110592
# labels	3	4	4	4	8	12	12	12	12	16
mean # iter.	7.27	7.9	8.05	10.79	31.85	49.1	49.4	49.4	49.7	66.1
<i>a priori</i> $C - 1$	1.	1.	1.	1.	1.	1.	1.	1.	1.	2.6332
<i>a posteriori</i>										
- first-max	0.0007	0.0231	0.2360	0.0030	0.0099	0.0102	0.0090	0.0101	0.0183	0.0209
- prob. mean	0.0010	0.0314	0.1073	0.0045	0.0177	0.0195	0.0174	0.0219	0.0309	0.0487
- prob. best	0.0007	0.0231	0.0547	0.0029	0.0138	0.0152	0.0134	0.0155	0.0247	0.0292

# Summary

- ▶ **Setting:**
  - ▶ Tight convex relaxation of multiclass image labeling
- ▶ **Bounds:**
  - ▶ *Probabilistic a priori* bound
  - ▶ Approximate generalized coarea formula
  - ▶ Compatible with finite-dimensional results



$$\mathcal{C}f(u) \geq \int_{\Gamma} f(\bar{u}_{\gamma}) d\mu(\gamma) = \mathbb{E}_{\gamma} f(\bar{u}_{\gamma})$$

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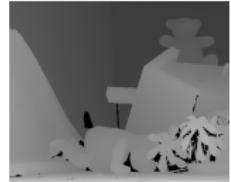
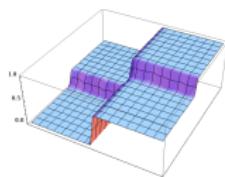
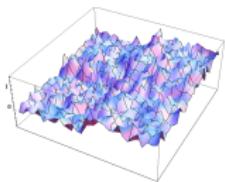
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SIAM IS12, May 2012

# Motivation – Multiclass Labeling

- ▶ **Applications:** Denoising, segmentation, 3D reconstruction, depth from stereo, inpainting, photo montage, optical flow,...



# Bounds – A priori: Properties

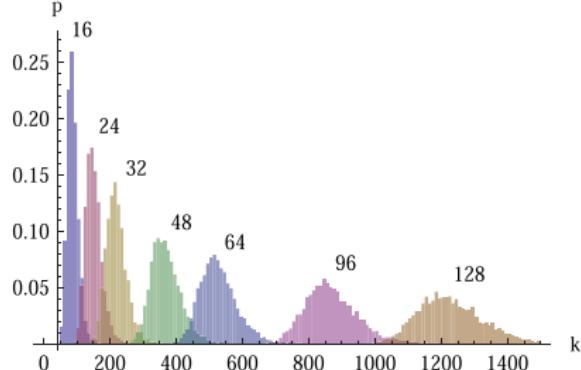
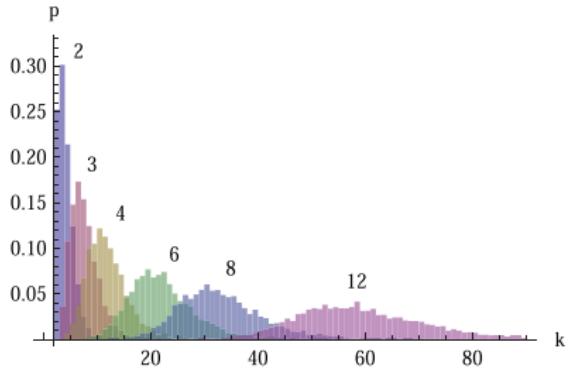
## Definition

For some sequence  $(\gamma^k)$ , if  $(u_\gamma^k)$  becomes stationary at some  $u_\gamma^{k'} \in \mathbb{N}$ , denote output  $\bar{u}_\gamma := u_\gamma^{k'}$ . For some functional  $f : \text{BV}(\Omega)^L \rightarrow \mathbb{R}$ , define

$$f(\bar{u}_\gamma) : \Gamma^\mathbb{N} \rightarrow \mathbb{R} \cup \{+\infty\}$$

$$\gamma \in \Gamma^\mathbb{N} \mapsto f(\bar{u}_\gamma) := \begin{cases} f(u_\gamma^{k'}), & (u_\gamma^k) \text{ stationary at } u_\gamma^{k'} \in \text{BV}(\Omega)^L, \\ +\infty, & \text{otherwise.} \end{cases} \quad (*)$$

# Experiments – Iterations



# Experiments II

