

# Uniqueness of the limiting profile for monotone Lipschitz random surfaces

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14 February 2020

Introduction

What is known

Main results

Moats

Potential class

Open problems

## Introduction

What is known

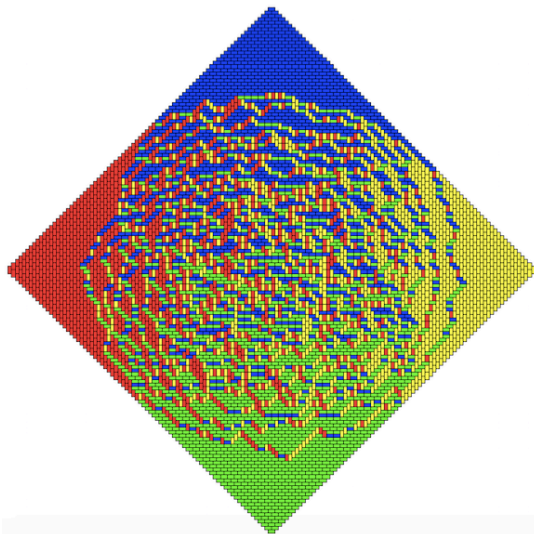
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# The variational principle for domino tilings



Uniformly random domino tiling of an aztec diamond

## The variational principle for domino tilings

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# The variational principle for domino tilings

$$Z_{\Lambda}^{\omega} := -\log \left( \begin{array}{c} \text{number of tilings of } \Lambda \text{ with} \\ \text{boundary height function} \\ \omega|_{\partial\Lambda} \end{array} \right)$$

Theorem (Cohn, Kenyon, Propp, 2000)

*Consider  $R$  an open region,  $\xi$  a continuous function on  $\partial R$ .*

*Assume:  $\frac{1}{n}\Lambda_n \rightarrow R$  and  $\frac{1}{n}\text{Graph}(\omega_n|_{\partial\Lambda_n}) \rightarrow \text{Graph}(\xi)$ ,*

*Then:  $\frac{1}{n^2}Z_{\Lambda_n}^{\omega_n} \rightarrow \inf_{f: f|_{\partial R}=\xi} \int_R \sigma(\nabla f) d\lambda$ ,*

*where  $\sigma(s)$  encodes the specific free energy of a gradient Gibbs measure of slope  $s$ .*

## The variational principle for domino tilings

If  $\sigma$  is strictly convex on the interior of the set of possible slopes, then there is a unique minimizer  $f_{\min}$  in the infimum

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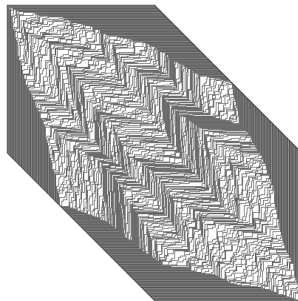
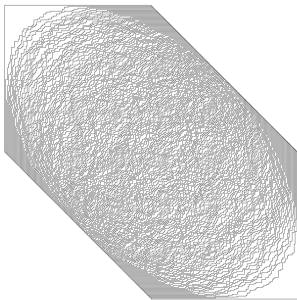
1. Uniqueness of the asymptotic profile,
2. The LDP with speed  $n^2$  and rate function

$$I(f) := \int_R \sigma(\nabla f) d\lambda$$

has a unique minimizer.

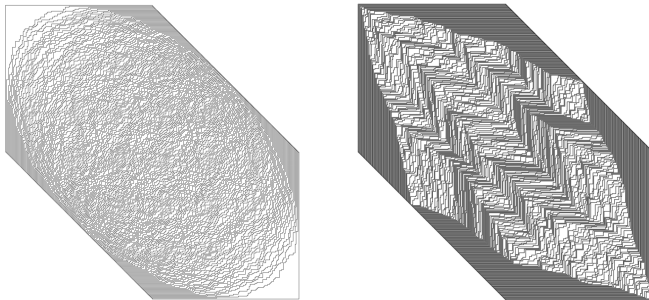
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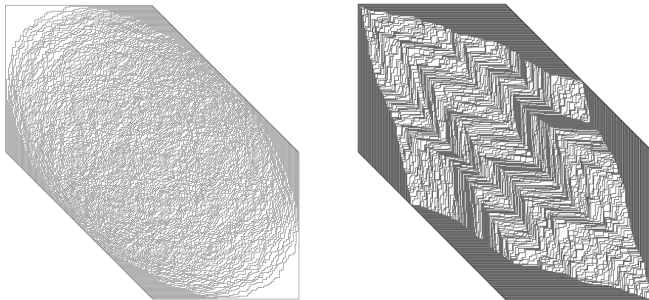


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## Goals:

1. Understand natural conditions that imply strict convexity,
2. Prove strict convexity for a large class of models.

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- ▶  $\phi$  is generated by a potential  $\Phi$ .

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Sheffield (2005) showed that  $\sigma$  is strictly convex for simply attractive potentials  $\Phi$ . These are potentials which are both:

1. Nearest neighbor: interactions between pairs of points only,
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4. Discrete Gaussian free field.

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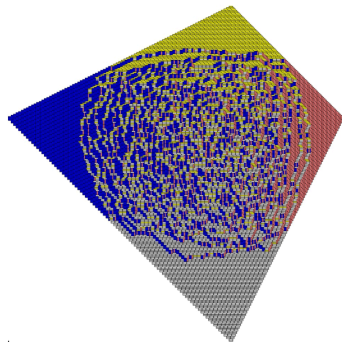
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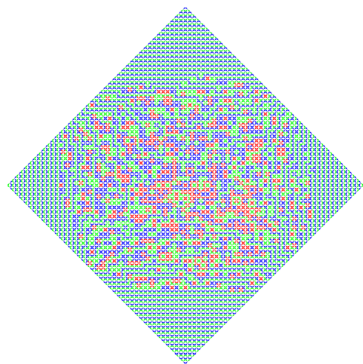
Open problems

## Can we go further?

Limit shapes appear for models with non-local interactions.



tiling by  $3 \times 1$  bars



tree-valued function

## Can we go further?

In order to attack more general random surfaces with more complex interactions we need to:

- ▶ Get around cluster swapping techniques which use edge energy and thus only works for two-points interactions,

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- ▶ Get around cluster swapping techniques which use edge energy and thus only works for two-points interactions,
- ▶ Understand what convexity of the potential is really bringing to the table.

The answer to these problems is **stochastic monotonicity**.

# Definition of stochastic monotonicity

## Definition

A specification  $\phi$  is **stochastically monotone** if

$$\phi_{\Lambda}^{\omega_1} \preceq \phi_{\Lambda}^{\omega_2} \quad \text{whenever} \quad \omega_1 \leq \omega_2,$$

where  $\omega_1, \omega_2$  are height functions,  $\Lambda \subset \subset \mathbb{Z}^d$ .



# Definition of Lipschitz

## Definition

A specification  $\phi$  is **Lipschitz** if there is a  $K < \infty$  such that  $\phi_\Lambda^\omega$  is supported on  $K$ -Lipschitz functions for  $\omega$   $K$ -Lipschitz.

# Statement of the result

## Theorem (L, Tassy)

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*Write  $U$  for the interior of  $\{\sigma < \infty\} \subset (\mathbb{R}^d)^*$ : the set of allowable slopes. The surface tension  $\sigma : U \rightarrow \mathbb{R}$  is strictly convex if:*

- 1.  $E = \mathbb{R}$*
- 2.  $E = \mathbb{Z}$  and  $\sigma$  is affine on  $\partial U$ , but not  $U$ .*

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$$|H_{\Lambda_n} - H_{\Lambda_n}^0| = o(n^d)$$

as  $n \rightarrow \infty$  where  $\Lambda_n = \{0, \dots, n-1\}^d \subset \mathbb{Z}^d$ .

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\* By which we mean: 1-Lipschitz w.r.t. some quasimetric  $q$ , in order to be as general as possible.

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- ▶ Can we extend the argument to non-Lipschitz functions?
- ▶ Is it possible to reformulate the variational principle in terms of specifications only?
- ▶ Can we find models non-monotone models for which there exists a  $c < \infty$  such that  $\omega_1 \leq \omega_2$  implies

$$\phi_{\Lambda}^{\omega_1} \preceq \phi_{\Lambda}^{\omega_2+c}?$$

If it is the case the Moat Lemma still works and  $\sigma$  is strictly convex.

