# Approximation via Correlation Decay when Strong Spatial Mixing Fails

Heng Guo

Queen Mary, University of London

Joint work with Ivona Bezáková, Andreas Galanis, Leslie Ann Goldberg, and Daniel Štefankovič

Rome, Italy

Jul 13 2016

1/23

## **Independent sets**

Graph 
$$G = (V, E)$$
.

An independent set is a subset of *V* such that no two are adjacent.

 $\mathfrak{I}(G)$  = the set of independent sets of G.

We are interested in approximating the size of  $\mathfrak{I}(G)$ .

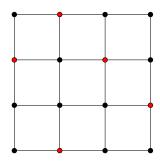
## **Independent sets**

Graph G = (V, E).

An independent set is a subset of *V* such that no two are adjacent.

 $\mathfrak{I}(G)$  = the set of independent sets of G.

We are interested in approximating the size of  $\mathfrak{I}(G)$ .



Independent

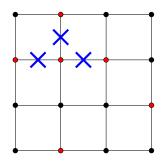
## **Independent sets**

Graph G = (V, E).

An independent set is a subset of *V* such that no two are adjacent.

 $\mathfrak{I}(G)$  = the set of independent sets of G.

We are interested in approximating the size of  $\mathfrak{I}(G)$ .



Not independent

#### The hardcore model

Alternatively, consider the following distribution:

$$\pi(\textit{I}) \propto \lambda^{|\textit{I}|}$$

for  $I \in \mathfrak{I}(G)$  and some  $\lambda > 0$ .

This is also called the hardcore model with activity  $\lambda$ .

#### The hardcore model

Alternatively, consider the following distribution:

$$\pi(I) \propto \lambda^{|I|}$$

for  $I \in \mathfrak{I}(G)$  and some  $\lambda > 0$ .

This is also called the hardcore model with activity  $\lambda$ .

Partition function  $Z = \sum_{I \in \mathfrak{I}(G)} \lambda^{|I|}$ .

In particular, if  $\lambda = 1$ ,  $Z = |\mathfrak{I}(G)|$ .

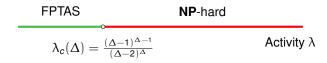
Approximate counting weighted independent sets (or approximate *Z* for the Hardcore model)

**ICALP 2016** 

4/23

Approximate counting weighted independent sets

(or approximate Z for the Hardcore model)



Approximate counting weighted independent sets (or approximate *Z* for the Hardcore model)

For G with a degree bound  $\Delta$ :

FPTAS NP-hard 
$$\lambda_c(\Delta) = \frac{(\Delta-1)^{\Delta-1}}{(\Delta-2)^{\Delta}} \qquad \qquad \text{Activity } \lambda$$

Algorithm: [Weitz 06]

Approximate counting weighted independent sets

(or approximate Z for the Hardcore model)

FPTAS NP-hard 
$$\lambda_c(\Delta) = \frac{(\Delta-1)^{\Delta-1}}{(\Delta-2)^{\Delta}} \qquad \qquad \text{Activity } \lambda$$

- Algorithm: [Weitz 06]
- Hardness: [Sly 10] [Sly Sun 14] [Galanis, Štefankovič, Vigoda 16]

Approximate counting weighted independent sets

(or approximate Z for the Hardcore model)

FPTAS NP-hard 
$$\lambda_c(\Delta) = \frac{(\Delta-1)^{\Delta-1}}{(\Delta-2)^{\Delta}} \qquad \qquad \text{Activity } \lambda$$

- Algorithm: [Weitz 06]
- Hardness: [Sly 10] [Sly Sun 14] [Galanis, Štefankovič, Vigoda 16]

## **Counting independent sets**

Specialize to approximate counting independent sets (fix  $\lambda = 1$ ):

FPTAS	<b>NP</b> -hard	
$\Delta \leqslant 5$	$\Delta\geqslant 6$	
( $\Delta=5$ is the largest integer so that $\lambda_c(\Delta)\geqslant 1)$		

- Algorithm: [Weitz 06]
- Hardness: [Sly 10]

Hypergraph H = (V, F), where a hyperedge  $e \in F$  is a subset of V.

Independent set  $I: I \subseteq V$  and  $\forall e \in F, e \not\subset I$ . (Not-All-IN)

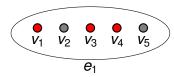
**ICALP 2016** 

6/23

Hypergraph H = (V, F), where a hyperedge  $e \in F$  is a subset of V.

Independent set  $I: I \subseteq V$  and  $\forall e \in F, e \not\subset I$ . (Not-All-IN)

#### Examples:

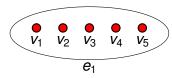


Independent

Hypergraph H = (V, F), where a hyperedge  $e \in F$  is a subset of V.

Independent set  $I: I \subseteq V$  and  $\forall e \in F, e \not\subset I$ . (Not-All-IN)

#### Examples:

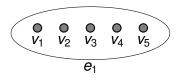


Not independent

Hypergraph H = (V, F), where a hyperedge  $e \in F$  is a subset of V.

Independent set  $I: I \subseteq V$  and  $\forall e \in F, e \not\subset I$ . (Not-All-IN)

## Examples:

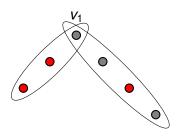


 $2^5 - 1$  many independent sets

Hypergraph H = (V, F), where a hyperedge  $e \in F$  is a subset of V.

Independent set  $I: I \subseteq V$  and  $\forall e \in F, e \not\subset I$ . (Not-All-IN)

#### Examples:

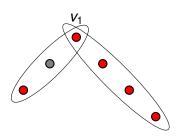


Independent

Hypergraph H = (V, F), where a hyperedge  $e \in F$  is a subset of V.

Independent set  $I: I \subseteq V$  and  $\forall e \in F, e \not\subset I$ . (Not-All-IN)

#### Examples:

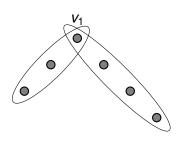


Not independent

Hypergraph H = (V, F), where a hyperedge  $e \in F$  is a subset of V.

Independent set  $I: I \subseteq V$  and  $\forall e \in F, e \not\subset I$ . (Not-All-IN)

#### Examples:



$$2^2 \cdot 2^3 + (2^2 - 1)(2^3 - 1)$$

many independent sets

#### **Monotone CNF**

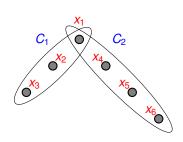
 $\label{eq:setsin} \mbox{Independent Sets in Hypergraphs} \Leftrightarrow \\ \mbox{Satisfying assignments of monotone CNF formulas}.$ 

7/23

#### **Monotone CNF**

 $\label{eq:setsin} \mbox{Independent Sets in Hypergraphs} \Leftrightarrow \\ \mbox{Satisfying assignments of monotone CNF formulas}.$ 

Vertices are variables. Hyperedges are clauses.



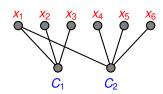
$$(\underbrace{x_1 \vee x_2 \vee x_3}_{C_1}) \wedge (\underbrace{x_1 \vee x_4 \vee x_5 \vee x_6}_{C_2})$$

#### **Monotone CNF**

# Independent Sets in Hypergraphs $\Leftrightarrow$

Satisfying assignments of monotone CNF formulas.

Vertices are variables. Hyperedges are clauses.



$$(\underbrace{x_1 \lor x_2 \lor x_3}_{C_1}) \land (\underbrace{x_1 \lor x_4 \lor x_5 \lor x_6}_{C_2})$$

#### **Bounded occurrences**

Name #HYPERINDSET( $\Delta$ , k).

Instance A hypergraph H with maximum degree at most  $\Delta$  where each hyperedge has cardinality (arity) at least k.

*Output* The number  $Z_H$  of independent sets in H.

**ICALP 2016** 

8/23

# **Previously**

#### Based on Markov chain Monte Carlo:

• There is a FPRAS for #HYPERINDSET( $\Delta$ , k) if  $k \ge \Delta + 2$ . [Dyer Greenhill 00], [Borderwich, Dyer, Karpinski 06, 08].

9/23

# **Previously**

#### Based on Markov chain Monte Carlo:

• There is a FPRAS for #HYPERINDSET( $\Delta$ , k) if  $k \ge \Delta + 2$ . [Dyer Greenhill 00], [Borderwich, Dyer, Karpinski 06, 08].

#### Based on correlation decay:

• There is a FPTAS for #HYPERINDSET( $\Delta, k$ ) if  $\Delta \leqslant 5$  for any integer  $k \geqslant 2$ . [Liu Lu 15]

# **Previously**

#### Based on Markov chain Monte Carlo:

• There is a FPRAS for #HYPERINDSET( $\Delta$ , k) if  $k \ge \Delta + 2$ . [Dyer Greenhill 00], [Borderwich, Dyer, Karpinski 06, 08].

#### Based on correlation decay:

• There is a FPTAS for #HYPERINDSET( $\Delta, k$ ) if  $\Delta \leqslant 5$  for any integer  $k \geqslant 2$ . [Liu Lu 15]

#HYPERINDSET( $\Delta$ , 2) is at least as hard as counting independent sets.

Hence FPTAS for k = 2,  $\Delta \leq 5$  is optimal. ( $\Delta \geq 6$  is **NP**-hard [Sly 10].)

#### **Our results**

#### **Theorem**

There is a FPTAS for  $\# HYPERINDSET(\Delta, k)$  if

- 2 For  $\Delta \geqslant 200$ ,  $k \geqslant 1.66\Delta$ .

#### **Our results**

#### **Theorem**

There is a FPTAS for  $\#HYPERINDSET(\Delta, k)$  if

- 2 For  $\Delta \ge 200$ ,  $k \ge 1.66 \Delta$   $k \ge \Delta$ .

#### Our results

#### **Theorem**

There is a FPTAS for #HYPERINDSET $(\Delta, k)$  if

- $\bullet$   $\Delta = 6$  and  $k \geqslant 3$ ;
- 2 For  $\Delta \geq 200$ ,  $k \geq 1.66\Delta$   $k \geq \Delta$ .

- For  $\Delta = 6$ , k = 3 is optimal as #HYPERINDSET(6, 2) is **NP**-hard [Sly 10].
- $k \ge \Delta$  only slightly improves  $k \ge \Delta + 2$  [Borderwich, Dyer, Karpinski 08], but this improvement is essential for our application of counting dominating sets in regular graphs.

#### **Hardness**

#### **Theorem**

For any integer  $\Delta \geqslant 5 \cdot 2^{k/2}$ , it is **NP**-hard to approximate #HYPERINDSET( $\Delta, k$ ), even within an exponential factor.

FPTAS	<b>NP</b> -hard
$\Delta \leqslant k$	$\Delta \geqslant 5 \cdot 2^{k/2}$

# **Dominating sets**

Graph G = (V, E).

 $D \subseteq V$  is dominating if every  $v \in V$  either  $\in D$  or is adjacent to some  $v' \in D$ .

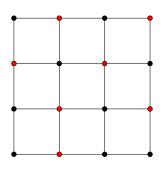
ICALP 2016

12 / 23

# **Dominating sets**

Graph G = (V, E).

 $D \subseteq V$  is dominating if every  $v \in V$  either  $\in D$  or is adjacent to some  $v' \in D$ .

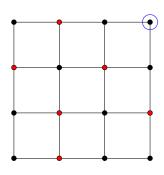


Dominating

# **Dominating sets**

Graph G = (V, E).

 $D \subseteq V$  is dominating if every  $v \in V$  either  $\in D$  or is adjacent to some  $v' \in D$ .



Not dominating

## **Counting dominating sets**

*Name*  $\#RegDomSet(\Delta)$ .

*Instance* A  $\triangle$ -regular graph G.

*Output* The number of dominating sets in *G*.

## Dominating sets $\leq_T$ Independent sets in hypergraphs

Express dominating sets as a CSP problem:

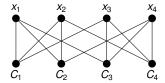
Each vertex is a variable and a constraint.

# **Dominating sets** $\leq_T$ **Independent sets in hypergraphs**

Express dominating sets as a CSP problem:

Each vertex is a variable and a constraint.



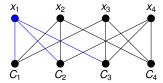


#### **Dominating sets** $\leq_T$ **Independent sets in hypergraphs**

Express dominating sets as a CSP problem:

Each vertex is a variable and a constraint.



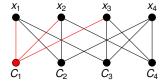


#### **Dominating sets** $\leq_T$ **Independent sets in hypergraphs**

Express dominating sets as a CSP problem:

Each vertex is a variable and a constraint.



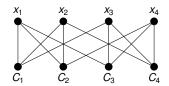


## **Dominating sets** $\leq_T$ **Independent sets in hypergraphs**

Express dominating sets as a CSP problem:

Each vertex is a variable and a constraint.





 $\#RegDomSet(\Delta) \leqslant_T \#HyperIndSet(\Delta + 1, \Delta + 1)$ 

$$\#RegDomSet(\Delta) \leqslant_T \#HyperIndSet(\Delta + 1, \Delta + 1)$$

15 / 23

$$\#RegDomSet(\Delta) \leqslant_T \#HyperIndSet(\Delta + 1, \Delta + 1)$$

#### **Theorem**

There is a FPTAS for #HYPERINDSET $(\Delta, k)$  if

- ② For  $\Delta \geqslant 200$ ,  $k \geqslant \Delta$ .

$$\#RegDomSet(\Delta) \leqslant_T \#HyperIndSet(\Delta + 1, \Delta + 1)$$

#### **Theorem**

There is a FPTAS for #HYPERINDSET $(\Delta, k)$  if

- 2 For  $\Delta \geqslant 200$ ,  $k \geqslant \Delta$ .

## Corollary

There is a FPTAS for  $\#REGDOMSET(\Delta)$  if  $\Delta \leqslant 5$  or  $\Delta \geqslant 199$ .

$$\#RegDomSet(\Delta) \leqslant_T \#HyperIndSet(\Delta + 1, \Delta + 1)$$

#### **Theorem**

There is a FPTAS for #HYPERINDSET $(\Delta, k)$  if

- 2 For  $\Delta \geqslant 200$ ,  $k \geqslant \Delta$ .

#### Corollary

There is a FPTAS for  $\#RegDomSet(\Delta)$  if  $\Delta \leq 5$  or  $\Delta \geq 199$ .

#### **Theorem**

Approximately counting dominating sets is **NP**-hard in graphs with degree bound  $\Delta \geqslant 18$ , even within an exponential factor.

# The Algorithm

## A recursion for counting independent sets

$$P_{G}(v) = \frac{|\{I \in \mathcal{I} \mid v \notin I\}|}{|\mathcal{I}|} = \frac{Z(G - v)}{Z(G)}$$

$$= \frac{Z(G - v)}{Z(G - v) + Z(G - v - N(v))}$$

$$= \frac{1}{1 + \frac{Z(G - v - N(v))}{Z(G - v)}}$$

## A recursion for counting independent sets

$$P_{G}(v) = \frac{|\{I \in \mathcal{I} \mid v \notin I\}|}{|\mathcal{I}|} = \frac{Z(G - v)}{Z(G)}$$

$$= \frac{Z(G - v)}{Z(G - v) + Z(G - v - N(v))}$$

$$= \frac{1}{1 + \frac{Z(G - v - N(v))}{Z(G - v)}}$$

Suppose  $N(v) = \{v_1, \ldots, v_d\}$ .

$$\frac{Z(G-v-N(v))}{Z(G-v)} = \frac{Z(G-v-v_1)}{Z(G-v)} \cdot \frac{Z(G-v-v_1-v_2)}{Z(G-v-v_1)} \cdot \dots \cdot \frac{Z(G-v-N(v))}{Z(G-v-(N(v)-v_d))}$$

## A recursion for counting independent sets

$$P_{G}(v) = \frac{|\{I \in \mathfrak{I} \mid v \in I\}|}{|\mathfrak{I}|} = \frac{Z(G - v)}{Z(G)}$$

$$= \frac{Z(G - v)}{Z(G - v) + Z(G - v - N(v))}$$

$$= \frac{1}{1 + \frac{Z(G - v - N(v))}{Z(G - v)}}$$

Suppose  $N(v) = \{v_1, \dots, v_d\}.$ 

$$\frac{Z(G-v-N(v))}{Z(G-v)} = \frac{Z(G-v-v_1)}{Z(G-v)} \cdot \frac{Z(G-v-v_1-v_2)}{Z(G-v-v_1)} \cdot \dots \cdot \frac{Z(G-v-N(v))}{Z(G-v-(N(v)-v_d))}$$

$$= P_{G_1}(v_1) \cdot P_{G_2}(v_2) \cdot \dots \cdot P_{G_d}(v_d)$$

Here  $G_i = G - v - v_1 - \cdots - v_{i-1}$ .

#### **Computation Tree**

The algorithm for #HYPERINDSET( $\Delta$ , k) is a similar recursion [Liu Lu 15].

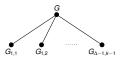
Except that  $\nu$  have  $(\Delta - 1)(k - 1)$  many neighbours.

## **Computation Tree**

The algorithm for #HYPERINDSET( $\Delta$ , k) is a similar recursion [Liu Lu 15].

Except that  $\nu$  have  $(\Delta - 1)(k - 1)$  many neighbours.

- The recursion forms a computation tree.
- We stop the recursion after  $O(\log n)$  many steps.
- The main task is to bound the error.



## Strong spatial mixing

SSM: Let  $\sigma_{\Lambda}$  and  $\tau_{\Lambda}$  be two partial configurations on  $\Lambda \subseteq V$ .

Let 
$$S$$
 be the set where  $\sigma_{\Lambda}$  and  $\tau_{\Lambda}$  differ.

$$|p_{\nu}^{\sigma_{\Lambda}} - p_{\nu}^{\tau_{\Lambda}}| \leqslant \exp(-\Omega(\operatorname{dist}(\nu, S)))$$

Roughly speaking, the influence of the boundary decays exponentially, even with some vertices fixed within the radius.

#### **SSM** for hypergraph independent sets

In the computation tree, hyperedge sizes will decrease.

Eventually, the size may go down to 2.

Hence SSM does not hold for independent sets in hypergraphs when  $\Delta \geqslant 6$ .

(SSM does not hold for independent sets in graphs if  $\Delta \geqslant 6$ .)

This is why [Liu, Lu 15] can only do  $\Delta \leqslant 5$ .

## Beyond strong spatial mixing

Our contribution is to provide a way to analyze correlation decay beyond the strong spatial mixing bound.

- Larger hyperedges have better decay.
- Keep track of the total "deficits" of sub-instances.
- Amortized analysis SSM is worst case.

## Main difficulty — bound the decay rate

Technically, the main difficulty is to bound the decay rate function — a optimisation problem with  $(\Delta - 1)(k - 1)$  many variables.

# Main difficulty — bound the decay rate

Technically, the main difficulty is to bound the decay rate function — a optimisation problem with  $(\Delta - 1)(k - 1)$  many variables.

#### **Decay Rate**

$$\kappa^{\textit{d},\textbf{k}}(\textbf{r}) := \frac{1}{\psi - \textit{F}(\textbf{r})^{\chi}} \sum_{i=1}^{\textit{d}} \alpha^{-\textit{l}_{k_i-1}} \frac{\prod_{j=1}^{k_i-1} \frac{\textit{r}_{i,j}}{1 + \textit{r}_{i,j}}}{1 - \prod_{j=1}^{k_i-1} \frac{\textit{r}_{i,j}}{1 + \textit{r}_{i,j}}} \sum_{j=1}^{k_i-1} \delta^{\textit{c}_{i,j}} \frac{\psi - \textit{r}_{i,j}^{\chi}}{1 + \textit{r}_{i,j}},$$

where

$$F(\mathbf{r}) = \prod_{i=1}^{d} \left( 1 - \prod_{j=1}^{k_i - 1} \frac{r_{i,j}}{1 + r_{i,j}} \right)$$

$$c_{i,j} = b_2(k-2) + s_{\min(i,d-b_2)} - \max(0,b_k'-i) - (j-1)(\Delta-1)\mathbf{1}_{i\leqslant d-b_2}.$$

#### **Open questions**

- The exact threshold for #HYPERINDSET( $\Delta$ , k)?
- Close the gap for  $\#REGDOMSET(\Delta)$ .
- Other instances where SSM fails to capture the complexity?

#### **Open questions**

- The exact threshold for #HYPERINDSET( $\Delta$ , k)?
- Close the gap for  $\#REGDOMSET(\Delta)$ .
- Other instances where SSM fails to capture the complexity?

# Thank You!

Full version: arxiv.org/abs/1510.09193