

# The complexity of choosing an $H$ -colouring (nearly) uniformly at random\*

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## Abstract

Cooper, Dyer and Frieze studied the problem of sampling  $H$ -colourings (nearly) uniformly at random. Special cases of this problem include sampling colourings and independent sets and sampling from statistical physics models such as the Widom-Rowlinson model, the Beach model, the Potts model and the hard-core lattice gas model. Cooper et al. considered the family of “cautious” ergodic Markov chains with uniform stationary distribution and showed that, for every fixed connected “nontrivial” graph  $H$ , every such chain mixes slowly. In this paper, we give a complexity result for the problem. Namely, we show that for **any** fixed graph  $H$  with no trivial components, there is unlikely to be any *Polynomial Almost Uniform Sampler* (PAUS) for  $H$ -colourings. We show that if there were a PAUS for the  $H$ -colouring problem, there would also be a PAUS for sampling independent sets in bipartite graphs and, by the self-reducibility of the latter problem, there would be a *Fully-Polynomial Randomised Approximation*

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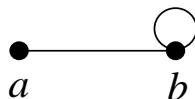


Figure 1: Homomorphisms from  $G$  to this graph are independent sets of  $G$ .

*Scheme* (FPRAS) for  $\#\text{BIS}$  — the problem of counting independent sets in bipartite graphs. Dyer, Goldberg, Greenhill and Jerrum have shown that  $\#\text{BIS}$  is complete in a certain logically-defined complexity class. Thus, a PAUS for sampling  $H$ -colourings would give an FPRAS for the entire complexity class. In order to achieve our result we introduce the new notion of *sampling-preserving* reduction which seems to be more useful in certain settings than approximation-preserving reduction.

## 1 Introduction

Let  $H = (V(H), E(H))$  be any fixed graph. An  $H$ -colouring of a graph  $G = (V(G), E(G))$  is just a homomorphism from  $G$  to  $H$ : The vertices of  $H$  correspond to “colours” and the edges of  $H$  specify which colours may be adjacent. Thus, an  $H$ -colouring of  $G$  is a function  $C$  from  $V(G)$  to  $V(H)$  such that for every edge  $(u, v) \in E(G)$ , the corresponding edge  $(C(u), C(v))$  is in  $E(H)$ . Informally, colours  $C(u)$  and  $C(v)$  are allowed to be adjacent in the colouring  $C$  of  $G$  because the edge  $(C(u), C(v))$  is an edge of  $H$ .

Many combinatorial problems can be viewed as special cases of  $H$ -colouring. For example, if  $H$  is a  $k$ -clique with no self-loops then  $H$ -colourings of  $G$  correspond to proper  $k$ -colourings of  $G$ . (In such a colouring,  $k$  colours are available for colouring the vertices of  $G$ , but no colour may be adjacent to itself.) Here is another example. If  $H$  is the graph depicted in Figure 1 then  $H$ -colourings of  $G$  correspond to independent sets of  $G$  — vertices which are coloured “ $a$ ” are in the independent set, and vertices which are coloured “ $b$ ” are not. Several models from statistical physics are special cases of  $H$ -colouring including the Widom-Rowlinson model, the Beach model, and (for weighted  $H$ -colourings) the Potts model and the hard-core lattice gas model. See [2, 10] for details.

The complexity of  $H$ -colouring has been well-studied. Many papers considered the following problem: Given a fixed graph  $H$ , determine, for an

input graph  $G$ , whether  $G$  has an  $H$ -colouring. Hell and Nešetřil [14] completely characterised the set of graphs for which this problem is NP-complete. They observed that the problem is in P if  $H$  has a loop or is bipartite and they showed that it is NP-complete for any other fixed  $H$ . See [14] for references to earlier work on this question and [13] for extensions to the case in which the maximum degree of  $G$  is bounded. See [4, 5] for extensions to parameterised complexity.

Dyer and Greenhill [10] considered the problem of *counting*  $H$ -colourings. Intriguingly, they were able to completely characterise the graphs  $H$  for which this problem is #P-complete. A connected component of  $H$  is said to be “trivial” if it is a complete graph with all loops present or a complete bipartite graph with no loops present<sup>1</sup>. Dyer and Greenhill showed that counting  $H$ -colourings is #P-complete if  $H$  has a nontrivial component and that it is in P otherwise. They also extended their result to the case in which the maximum degree of  $G$  is bounded.

Other work has focused on the complexity of *sampling*  $H$ -colourings (nearly) uniformly at random<sup>2</sup>. Positive results for particular graphs  $H$  (specifically for the case in which  $H$ -colourings are independent sets and for the case in which  $H$ -colourings are proper colourings) appear in works such as [9, 15, 18]. A negative result for the independent-set case appears in [6]. The first paper to study the complexity of sampling  $H$ -colourings in the general case was Cooper, Dyer and Frieze [3]. They focused on connected graphs  $H$  for which the decision problem “Is there an  $H$ -colouring?” is in P, but the counting problem “How many  $H$ -colourings are there?” is #P-complete. They showed that for any such  $H$ ,  $H$ -colourings cannot be sampled efficiently using “cautious” Markov chains, which are Markov chains which can change only a constant fraction of the colours of the vertices in a single step. In particular, the mixing time of all such chains is exponential in the number of vertices of  $G$ . They also give positive results for certain weighted cases, which are extended in [12]. In particular, [12] shows that for every fixed “dismantleable”  $H$  and every degree bound  $\Delta$ , there are positive vertex-weights

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<sup>1</sup>Following the usual notation in the area, we will treat self-loops specially, so it makes sense to refer to bipartite graphs with or without loops. The loop-free single-vertex is viewed as a complete bipartite graph.

<sup>2</sup>Some of this work has been motivated by the well-known connection between almost-uniform sampling and approximate counting [17, 8]. For some graphs  $H$ , it can be shown that the problem of approximately counting  $H$ -colourings is equivalent to the problem of sampling  $H$ -colourings (nearly) uniformly at random. See Section 8.

which can be assigned to the vertices of  $H$  so that weighted  $H$ -colourings can be sampled for degree- $\Delta$  graphs  $G$ . Borgs et al. [1] consider the problem of sampling  $H$ -colourings on rectangular subsets of the hypercubic lattice. They show that for every nontrivial connected  $H$  there is an assignments of weights to colours for which cautious chains are slowly mixing.

In this work, we study the complexity of sampling  $H$ -colourings. We show that if  $H$  has no trivial components then the problem of nearly-uniformly sampling  $H$ -colourings is intractable in a complexity-theoretic sense. In particular, we show that for any fixed  $H$  with no trivial components, there is unlikely to be any *Polynomial Almost Uniform Sampler* (PAUS) for  $H$ -colourings. We show that if there were a PAUS for the  $H$ -colouring problem, there would also be a PAUS for sampling independent sets in bipartite graphs and, by the self-reducibility of the latter problem, there would be a *Fully-Polynomial Randomised Approximation Scheme* (FPRAS) for  $\#BIS$  — the problem of counting independent sets in bipartite graphs. Dyer, Goldberg, Greenhill and Jerrum have shown that  $\#BIS$  is complete in a certain logically-defined subclass of  $\#P$  which includes problems such as counting downsets in partial orders and counting satisfying assignments in “restricted Horn” CNF Boolean formulas. Thus, a PAUS for sampling  $H$ -colourings would give an FPRAS for the entire complexity class. In fact, our result holds even if the input  $G$  is restricted to be a connected bipartite graph.

In order to achieve our result we introduce the new notion of *sampling-preserving* reduction. The notion of approximation-preserving reduction (AP-reducibility) from [11] seems to be too demanding. In particular, since AP-reducibility is about *counting* (as opposed to sampling), an AP-reduction is not allowed to inflate the size of the set of structures by a factor which is difficult to compute. Sampling-preserving reductions allow this flexibility while achieving the same final result. The definition of sampling reduction (Section 2) is essentially many-one. Nevertheless the reductions get used in a “Turing reduction” way. In particular, our reduction from  $SAMPLEBIS$  to  $SAMPLEH-COL$  takes an instance of  $SAMPLEBIS$  and constructs many  $SAMPLEH-COL$  instances. Since the resulting maps between  $H$ -colourings and independent sets are many-one, several reductions can be combined even though they may involve different amounts of inflation of the state space.

The paper is structured as follows. Section 2 gives the relevant definitions including the definition of a sampling-preserving reduction. Section 3 presents some technical lemmas which we need in our proofs. Section 4 outlines a general proof technique for demonstrating the existence of an SP-

reduction. Section 5 uses the new proof technique to reduce SAMPLEBIS to a crucial intermediate problem, SAMPLEFIXEDH-COL. Section 6 proves the main result. Sections 7 and 8 discuss extensions.

## 2 Definitions

The total variation distance between two distributions  $\pi$  and  $\pi'$  on a countable set  $\Omega$  is given by

$$d_{\text{TV}}(\pi, \pi') = \frac{1}{2} \sum_{\omega \in \Omega} |\pi(\omega) - \pi'(\omega)| = \max_{A \subseteq \Omega} |\pi(A) - \pi'(A)|.$$

A sampling problem  $X$  maps each instance  $\sigma$  to a set of structures  $X(\sigma)$ . The goal is to produce a member of  $X(\sigma)$  uniformly at random. The size of each structure in  $X(\sigma)$  is at most a polynomial in  $|\sigma|$ . For a given graph  $H$ , the sampling problem SAMPLEH-COL will be defined as follows.

*Name.* SAMPLEH-COL.

*Instance.* A loop-free graph  $G$ .

*Output.* An  $H$ -colouring of  $G$  chosen uniformly at random.

We will be particularly interested in the special case of this problem in which the input graph,  $G$ , is connected and bipartite.

*Name.* SAMPLEBH-COL.

*Instance.* A loop-free connected bipartite graph  $G$ .

*Output.* An  $H$ -colouring of  $G$  chosen uniformly at random.

The problem SAMPLEBIS will be defined as follows.

*Name.* SAMPLEBIS.

*Instance.* A loop-free connected bipartite graph  $G$ .

*Output.* An independent set of  $G$  chosen uniformly at random.

An *almost uniform sampler* [8, 16, 17] for  $X$  is a randomised algorithm that takes input  $\sigma$  and accuracy parameter  $\epsilon \in (0, 1]$  and gives an output such that the variation distance between the output distribution of the algorithm and the uniform distribution on  $X(\sigma)$  is at most  $\epsilon$ . We will say that

algorithm is a *polynomial almost uniform sampler (PAUS)* if its running time is bounded from above by a polynomial in the size of the instance  $|\sigma|$  and  $1/\epsilon$ .

A *sampling-preserving reduction* (SP-reduction) from a sampling problem  $X$  to a sampling problem  $Y$  (denoted  $X \leq_{\text{SP}} Y$ ) consists of

1. A function  $f$  which takes an input  $(\sigma, \epsilon)$ , in which  $\sigma$  is an instance of  $X$  and  $\epsilon \in (0, 1]$  is an accuracy parameter, and produces an instance  $f(\sigma, \epsilon)$  of  $Y$ . If  $X(\sigma)$  is non-empty then  $Y(f(\sigma, \epsilon))$  must be non-empty.
2. A function  $g$  which maps each tuple  $(\sigma, \epsilon, y)$  with  $y \in Y(f(\sigma, \epsilon))$  to a member of  $X(\sigma) \cup \{\perp\}$  where “ $\perp$ ” is an error symbol and for every  $(\sigma, \epsilon)$  and every  $x \in X(\sigma)$ ,

$$e^{-\epsilon} \frac{|Y(f(\sigma, \epsilon))|}{|X(\sigma)|} \leq |\{y \in Y(f(\sigma, \epsilon)) \mid g(\sigma, \epsilon, y) = x\}| \leq e^{\epsilon} \frac{|Y(f(\sigma, \epsilon))|}{|X(\sigma)|}. \quad (1)$$

Equation (1) says that for every  $x \in X(\sigma)$ , the number of  $y \in Y(f(\sigma, \epsilon))$  which are mapped to  $x$  by  $g$  is roughly  $\frac{|Y(f(\sigma, \epsilon))|}{|X(\sigma)|}$ . Thus, each  $x \in X(\sigma)$  is roughly equally represented and the error symbol  $\perp$  is represented by only about an  $\epsilon$ -fraction of  $Y(f(\sigma, \epsilon))$ .

The functions  $f$  and  $g$  must be computable in time which is bounded by a polynomial in  $|\sigma|$  and  $1/\epsilon$ .

The motivation for this definition is the following lemma.

**Lemma 1** *If  $X \leq_{\text{SP}} Y$  and  $Y$  has a PAUS, then  $X$  has a PAUS.*

*Proof.* Let  $(f, g)$  be the reduction from  $X$  to  $Y$  and let  $\mathcal{A}$  be a PAUS for  $Y$ . Here is a PAUS for  $X$ : On input  $(\sigma, \epsilon)$ , let  $y$  be the output of  $\mathcal{A}$  when it is run with inputs  $f(\sigma, \epsilon/4)$  and  $\epsilon/2$ ; return  $g(\sigma, \epsilon/4, y)$ . We must show that the variation distance between the output distribution of this algorithm and the uniform distribution on  $X(\sigma)$  is at most  $\epsilon$ . Let  $\sigma$  be an input with  $|X(\sigma)| \geq 1$ . Consider any subset  $A_x$  of  $X(\sigma)$ . Let

$$A_y = \{y \in Y(f(\sigma, \epsilon/4)) \mid g(\sigma, \epsilon/4, y) \in A_x\}.$$

Then the probability that  $\mathcal{A}$  gives an output in  $A_y$  is at most

$$\begin{aligned}
& \frac{|A_y|}{|Y(f(\sigma, \epsilon/4))|} + \frac{\epsilon}{2} \\
& \leq \frac{e^{\epsilon/4}|A_x|}{|X(\sigma)|} + \frac{\epsilon}{2} \\
& \leq \frac{(1 + \epsilon/2)|A_x|}{|X(\sigma)|} + \frac{\epsilon}{2} \\
& \leq \frac{|A_x|}{|X(\sigma)|} + \frac{(\epsilon/2)|A_x|}{|X(\sigma)|} + \frac{\epsilon}{2} \\
& \leq \frac{|A_x|}{|X(\sigma)|} + \epsilon.
\end{aligned}$$

Also, the probability that  $\mathcal{A}$  gives an output in  $A_y$  is at least

$$\begin{aligned}
& \frac{|A_y|}{|Y(f(\sigma, \epsilon/4))|} - \frac{\epsilon}{2} \\
& \geq \frac{e^{-\epsilon/4}|A_x|}{|X(\sigma)|} - \frac{\epsilon}{2} \\
& \geq \frac{(1 - \epsilon/2)|A_x|}{|X(\sigma)|} - \frac{\epsilon}{2} \\
& \geq \frac{|A_x|}{|X(\sigma)|} - \frac{|A_x|(\epsilon/2)}{|X(\sigma)|} - \frac{\epsilon}{2} \\
& \geq \frac{|A_x|}{|X(\sigma)|} - \epsilon.
\end{aligned}$$

□

The problem #BIS is defined as follows.

*Name.* #BIS.

*Instance.* A loop-free bipartite graph  $G$ .

*Output.* The number of independent sets of  $G$ .

A component of  $H$  is *trivial* if it is a complete graph with all loops present or a complete bipartite graph with no loops present. Recall from Dyer and Greenhill [10] that counting  $H$ -colourings is in P if  $H$  is trivial. The main result of this paper is as follows.

**Theorem 2** *Suppose that  $H$  is a fixed graph with no trivial components. If  $\text{SAMPLEBH-COL}$  has a PAUS then  $\text{SAMPLEBIS}$  has a PAUS and  $\#\text{BIS}$  has an FPRAS. Thus, every problem which is AP-interreducible with  $\#\text{BIS}$  (see [11]) has an FPRAS.*

### 3 Technical lemmas

Let  $\nu(a, b)$  denote the number of onto functions from a set of size  $a$  to a set of size  $b$ . We need to use the following lemma, which is Lemma 18 of [11].

**Lemma 3** (DGGJ) *If  $a$  and  $b$  are positive integers and  $a \geq 2b \ln b$  then*

$$b^a (1 - \exp(-a/(2b))) \leq \nu(a, b) \leq b^a.$$

We also need the following technical lemma.

**Lemma 4** *Suppose  $c_1$  and  $c_2$  are fixed positive reals with  $c_1 < c_2$ . For any  $\delta > 0$  and any non-negative integers  $q$  and  $a_0$ , there are non-negative integers  $a$  and  $b$  with  $a \geq a_0$  which are in  $O((a_0 + q)/\delta)$  and satisfy*

$$e^{-\delta} c_2^{a+q} \leq c_1^{b+q} \leq e^{\delta} c_2^{a+q}.$$

*Proof.* First, note that it would suffice to find non-negative integers  $a'$  and  $b'$  which are in  $O(q'/\delta)$  and satisfy

$$e^{-\delta} c_2^{a'+q'} \leq c_1^{b'+q'} \leq e^{\delta} c_2^{a'+q'},$$

where  $q' = q + a_0$  because we could simply set  $a = a' + a_0$  and  $b = b' + a_0$  which would imply  $a' + q' = a + q$  and  $b' + q' = b + q$ .

Taking logarithms, what we need is

$$\left| b' - \frac{a' \log c_2 + q' \log(c_2/c_1)}{\log c_1} \right| \leq \frac{\delta}{\log c_1}. \quad (2)$$

Now let  $\rho$  be defined by  $c_2 = c_1^{1+\rho}$ . Then we want

$$|b' - (a'(1 + \rho) + q'\rho)| \leq \frac{\delta}{\log c_1}. \quad (3)$$

For a positive integer  $r$ , we will choose  $a' = q'r$ , so we want

$$|b' - a' - \rho q'(r+1)| \leq \frac{\delta}{\log c_1}. \quad (4)$$

Let  $R = \lceil 2 \log c_1 / \delta \rceil$ . Lemma 19 of [11] says: *For any real  $z > 0$  and any positive integer  $R$  there is an  $x \in [1, \dots, R]$  such that*

$$\min(zx - \lfloor zx \rfloor, \lceil zx \rceil - zx) \leq 1/R.$$

Thus, there is an  $x \in [1, \dots, R]$  such that  $\rho q'x$  is within  $1/R$  of a non-negative integer. If  $x > 1$  we will set  $r+1 = x$ . If  $x = 1$  then note that  $\rho q'2$  is within  $2/R$  of a non-negative integer, so we will set  $r = 1$ .

Now recall that  $a' = q'r$ , so  $a' \in O(q'/\delta)$  as required.  $\square$

## 4 Demonstrating the existence of SP-reductions: a proof technique

When we introduce an SP-reduction from a sampling problem  $X$  to a sampling problem  $Y$ , we will need to show that Equation (1) is satisfied. We will typically do this by partitioning  $Y(f(\sigma, \epsilon))$  into disjoint sets  $Y_0, \dots, Y_k$ . We will show that each of  $Y_1, \dots, Y_k$  is fairly representative of  $X(\sigma)$ . In particular, for every  $x \in X(\sigma)$  and every  $i \in [1, k]$ ,

$$e^{-\epsilon/2} \frac{|Y_i|}{|X(\sigma)|} \leq |\{y \in Y_i \mid g(\sigma, \epsilon, y) = x\}| \leq e^{\epsilon/2} \frac{|Y_i|}{|X(\sigma)|}. \quad (5)$$

For every  $y \in Y_0$ , we will have  $g(\sigma, \epsilon, y) = \perp$  but we will show that  $Y_0$  is a small part of  $Y(f(\sigma, \epsilon))$ . In particular,

$$\sum_{i=1}^k |Y_i| \geq e^{-\epsilon/2} |Y(f(\sigma, \epsilon))|. \quad (6)$$

Together, (5) and (6) imply (1). Note that (6) follows from

$$|Y_0| \leq (\epsilon/4) |Y(f(\sigma, \epsilon))|, \quad (7)$$

since (7) implies  $|Y| - |Y_0| \geq (1 - \epsilon/4) |Y(f(\sigma, \epsilon))| \geq e^{-\epsilon/2} |Y(f(\sigma, \epsilon))|$ .

Quite often the reduction  $X \leq_{\text{SP}} Y$  will involve several subproblems  $Z_1, Z_2, \dots$  such that, for each of these, an SP-reduction  $(f_i, g_i)$  from  $X$  to  $Z_i$  is already known. The instance  $f(\sigma, \epsilon)$  of  $Y$  is then formed by “gluing” together instances  $f_1(\sigma, \epsilon/2)$  of  $Z_1$ ,  $f_2(\sigma, \epsilon/2)$  of  $Z_2$ , and so on.  $Y_i$  is (roughly) the portion of  $Y(f(\sigma, \epsilon))$  for which, within each  $y \in Y_i$ , we can “zoom in” on a structure  $z \in Z_i(f_i(\sigma, \epsilon/2))$ . Each structure in  $Z_i(f_i(\sigma, \epsilon/2))$  is represented by an equal number of  $y \in Y_i$  so we can get (5) by referring to the SP-reduction from  $X$  to  $Z_i$ . Establishing (7) is essentially showing that, although  $Y(f(\sigma, \epsilon))$  has some structures which don’t allow us to “zoom in” on an appropriate sub-problem to find our sample, these aren’t so numerous.

Finally, let  $Y_i(x) = \{y \in Y_i \mid g(\sigma, \epsilon, y) = x\}$ . Suppose that no  $y \in Y_i$  has  $g(\sigma, \epsilon, y) = \perp$ . In this case we can show (5) by showing that for all  $x, x' \in X(\sigma)$ ,

$$|Y_i(x)| \leq e^{\epsilon/2} |Y_i(x')|. \quad (8)$$

To see this, note that

$$\frac{|Y_i|}{|X(\sigma)|} = \frac{\sum_{x' \in X(\sigma)} |Y_i(x')|}{|X(\sigma)|} \geq e^{-\epsilon/2} \frac{\sum_{x' \in X(\sigma)} |Y_i(x')|}{|X(\sigma)|} = e^{-\epsilon/2} |Y_i(x)|.$$

## 5 Sampling *fixed* $H$ -colourings

Suppose that  $H$  is connected, loop-free, and bipartite. Denote the vertex partition of  $H$  by  $(V_L(H), V_R(H))$ . We will define the *fixed*  $H$ -colouring problem as follows.

*Name.* SAMPLEFIXED $H$ -COL

*Instance.* A loop-free connected bipartite graph  $G$  with vertex partition  $(V_L(G), V_R(G))$

*Output.* An  $H$ -colouring of  $G$  chosen uniformly at random from the set of  $H$ -colourings in which vertices of  $V_L(G)$  receive colours from  $V_L(H)$ .

We will study the problem SAMPLEFIXED $H$ -COL as an intermediate step on the way to the proof of Theorem 2.

A vertex in  $V_L(H)$  is said to be *full* if it is adjacent to every vertex in  $V_R(H)$ . Similarly, a vertex in  $V_R(H)$  is said to be *full* if it is adjacent to every vertex in  $V_L(H)$ . The graph  $H$  is said to be *full* if both  $V_L(H)$  and  $V_R(H)$  contain at least one full vertex. The following lemma is the key ingredient in the proof of Theorem 2.

**Lemma 5** *Suppose that  $H$  is a connected nontrivial full loop-free bipartite graph. Then  $\text{SAMPLEBIS} \leq_{\text{SP}} \text{SAMPLEFIXEDH-COL}$ .*

*Proof.* We'll prove the lemma by induction on the number of vertices in  $H$ . For the base case, suppose that  $H$  has at most 4 vertices. The only connected nontrivial full loop-free bipartite graph  $H$  with at most 4 vertices is the path of length 3. Let  $G$  be an input to  $\text{SAMPLEBIS}$ . There is a one-to-one correspondence between independent sets of  $G$  and fixed  $H$ -colourings of  $G$ : The endpoints of  $H$  point out the vertices which are in the independent set (see the proof of Theorem 5 of [11]).

We will now move on to the inductive step. The high-level idea is the following. By considering the graph  $H$ , we will construct several graphs  $H_{S_1}, \dots, H_{S_{j+k}}$ , each of which is smaller than  $H$  and satisfies certain conditions. By induction, for each  $i$ , there is an SP-reduction from  $\text{SAMPLEBIS}$  to  $\text{SAMPLEFIXEDH}_{S_i}\text{-COL}$ . If we apply this reduction to our instance  $G$  of  $\text{SAMPLEBIS}$ , we get an instance  $G_i$  of  $\text{SAMPLEFIXEDH}_{S_i}\text{-COL}$ . Our goal is to construct an instance  $f(G, \epsilon)$  of  $\text{SAMPLEFIXEDH-COL}$ . We do this by “gluing together” the various  $G_i$ 's. Now consider the constructed instance  $f(G, \epsilon)$  of  $\text{SAMPLEFIXEDH-COL}$ . When we sample from the output distribution  $\text{SAMPLEFIXEDH-COL}(f(G, \epsilon))$ , we would like to recover the output distribution of  $\text{SampleBIS}(G)$ . Curiously, we can not determine during the reduction itself the relative weights of the sub-instances  $G_1, G_2, \dots$ . Nevertheless, once we have an output to  $\text{SAMPLEFIXEDH-COL}(f(G, \epsilon))$ , the output itself tells us which  $H_i$  is relevant. From this, we can recover an output to  $\text{SAMPLEFIXEDH}_{S_i}\text{-COL}(G_i)$  and from this we can recover an output to  $\text{SAMPLEBIS}(G)$ . The main technical difficulty lies in showing that the distributions are correct. In particular, since the sub-reductions are SP-reductions (i.e., the equations in Section 4 are satisfied in the construction of  $G_1, G_2, \dots$ ), the combined reduction is also an SP-reduction.

We now describe the details. Let  $F_L$  be the set of full vertices in  $V_L(H)$  and let  $F_R$  be the set of full vertices in  $V_R(H)$ . Let  $f_L = |F_L|$  and  $f_R = |F_R|$  and  $v_L = |V_L(H)|$  and  $v_R = |V_R(H)|$ . For a subset  $S$  of  $V_R(H)$ , let  $N(S)$  be the set of mutual neighbours of  $S$ :

$$N(S) = \{v \in V_L(H) \mid \forall u \in S, (u, v) \in E(H)\}.$$

Note that  $F_L \subseteq N(S) \subseteq V_L(H)$ .  $S$  is said to be *left-reducing* if  $F_L \subset N(S) \subset V_L(H)$ . If  $S$  is left-reducing, let  $H_S$  be the subgraph of  $H$  induced by vertex partition  $(N(S), V_R(H))$ . Note that  $H_S$  has fewer vertices than  $H$ . Also, it

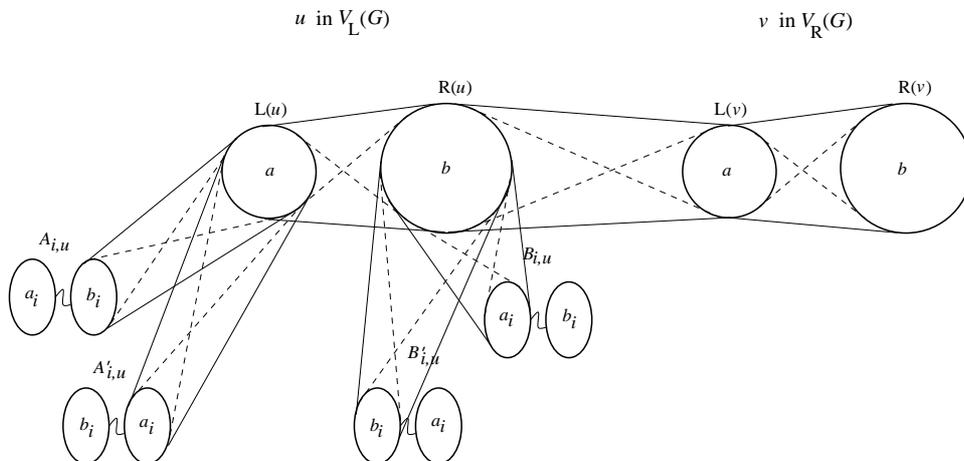


Figure 2: The construction of  $f(G, \epsilon)$  in the proof of Lemma 5.

is connected, full and nontrivial: The set of full vertices in  $V_L(H_S)$  is  $F_L$ ; the set of full vertices in  $V_R(H_S)$  includes all of  $F_R$  but it does not equal  $V_R(H)$  since  $N(S) \supset F_L$ .

Similarly, a subset  $S$  of  $V_L(H)$  is *right-reducing* if  $F_R \subset N(S) \subset V_R(H)$ . If  $S$  is right-reducing, let  $H_S$  be the subgraph of  $H$  induced by vertex partition  $(V_L(H), N(S))$ .  $H_S$  has fewer vertices than  $H$  and is connected, full and nontrivial.

Now, let  $S_1, \dots, S_k$  be the left-reducing subsets of  $V_R(H)$  and let  $S_{k+1}, \dots, S_{k+j}$  be the right-reducing subsets of  $V_L(H)$ . (Either or both of  $k$  and  $j$  could be zero.) For every  $i \in \{1, \dots, k+j\}$ , let  $(f_i, g_i)$  be an SP-reduction from SAMPLEBIS to SAMPLEFIXED $H_{S_i}$ -COL. Take the input  $(G, \epsilon)$  to SAMPLEBIS and let  $G_i = f_i(G, \epsilon/2)$ . Let  $a_i = |V_L(G_i)|$  and let  $b_i = |V_R(G_i)|$ . Let  $q = \sum_{i=1}^{k+j} (a_i + b_i)$  and let  $n = |V_L(G)| + |V_R(G)|$ .

Let  $f(G, \epsilon)$  be the graph which is constructed as follows, where  $a$  and  $b$  will be chosen later to satisfy

$$a \geq 2v_L \lceil q \ln(v_R/f_R) + \ln(16n/\epsilon) \rceil, \quad (9)$$

and

$$b \geq 2v_R \lceil q \ln(v_L/f_L) + \ln(16n/\epsilon) \rceil. \quad (10)$$

See Figure 2. For every vertex  $u$  of  $G$ , put a size- $a$  set  $L(u)$  into  $V_L(f(G, \epsilon))$  and a size- $b$  set  $R(u)$  into  $V_R(f(G, \epsilon))$ . Also, add edges  $L(u) \times R(u)$  to

$E(f(G, \epsilon))$ . If  $u \in V_L(G)$  is connected to  $v \in V_R(G)$  by an edge of  $G$  then add edges  $R(u) \times L(v)$  to  $E(f(G, \epsilon))$ .

Also, for every vertex  $u$  of  $G$  and every  $i \in [1, \dots, k+j]$ , let  $A_{i,u}$  and  $B_{i,u}$  be copies of  $G_i$  and let  $A'_{i,u}$  and  $B'_{i,u}$  be copies of  $G_i$  in which left-vertices and right-vertices are switched (so the vertices in  $V_L(A'_{i,u})$  correspond to the vertices in  $V_R(G_i)$  and the vertices in  $V_R(A'_{i,u})$  correspond to the vertices in  $V_L(G_i)$ ). Add edges  $L(u) \times V_R(A_{i,u})$  and  $L(u) \times V_R(A'_{i,u})$  and  $R(u) \times V_L(B_{i,u})$  and  $R(u) \times V_L(B'_{i,u})$  to  $E(f(G, \epsilon))$ .

Let

$$V_L(f(G, \epsilon)) = \bigcup_u L(u) \cup \bigcup_{u,i} \{V_L(A_{i,u}) \cup V_L(A'_{i,u}) \cup V_L(B_{i,u}) \cup V_L(B'_{i,u})\}$$

and let  $Y$  be the set of fixed  $H$ -colourings of  $f(G, \epsilon)$ . We will partition  $Y$  into sets  $Y_0, \dots, Y_{k+j+1}$ .

For  $i \in [1, \dots, k]$ ,  $Y_i$  is the set of colourings which are not in  $Y_1, \dots, Y_{i-1}$  but in which some  $u \in V_L(G)$  has  $R(u)$  coloured with (exactly) the colours in  $S_i$ . For  $i \in [k+1, \dots, k+j]$ ,  $Y_i$  is the set of colourings which are not in  $Y_1, \dots, Y_{i-1}$  but in which some  $v \in V_R(G)$  has  $L(v)$  coloured with  $S_i$ .

The high-level structure of our construction is as follows. For every  $i \in \{1, \dots, k+j\}$ , we will use the colourings in  $Y_i$  by focusing on the induced colourings of the subgraph  $G_i$ . These are  $H_{S_i}$ -colourings of  $G_i$  and from these we can (by induction) recover a random independent set of  $G$ . As usual, the colourings in  $Y_0$  are not useful for pointing out independent sets, but there are not too many of these. Every colouring in  $Y_{k+j+1}$  has a special form — Every vertex  $u$  of  $G$  either has  $R(u)$  coloured  $V_R(H)$  or has  $L(u)$  coloured  $V_L(H)$ . These colourings point out independent sets of  $G$  in a natural way, and each independent set comes up about the same number of times in this way.

We now look at the details. Note that for any colourings in  $Y_0$  or  $Y_{k+j+1}$ , we have the following property — every vertex  $u \in V_L(G)$  has  $R(u)$  coloured with a set  $S$  of colours such that  $N(S)$  is either  $F_L$  or  $V_L(H)$ . Similarly, every vertex  $v \in V_R(G)$  has  $L(v)$  coloured with a set  $S$  of colours such that  $N(S)$  is either  $F_R$  or  $V_R(H)$ .

Consider a colouring  $y$ . Vertex  $u \in V_L(G)$  satisfies *Condition (A)* if  $R(u)$  is coloured with a set  $S$  of colours with  $N(S) = F_L$  but  $S \subset V_R(H)$ . It satisfies *Condition (B)* if  $R(u)$  is coloured with a set  $S$  of colours with  $N(S) = V_L(H)$  but  $L(u)$  is coloured with a proper subset of  $V_L(H)$ . Vertex  $v \in V_R(G)$  satisfies *Condition (C)* if  $L(v)$  coloured with a set  $S$  of colours with  $N(S) =$

$F_R$  but  $S \subset V_L(H)$ . It satisfies *Condition (D)* if  $L(v)$  is coloured with a set  $S$  of colours with  $N(S) = V_R(H)$  but  $R(v)$  is coloured with a proper subset of  $V_R(H)$ .

We now define

$$Y_0 = \{y \in Y - \{Y_1 \cup \dots \cup Y_{k+j}\} \mid \text{some vertex satisfies Condition A or B or C or D}\}.$$

Now note that colourings in  $Y_{k+j+1}$  have the following property. Every vertex  $u$  of  $G$  either has  $R(u)$  coloured  $V_R(H)$  or has  $L(u)$  coloured  $V_L(H)$ .

We will first work on establishing Equation (7). Let  $Y_{u,A}$  denote the subset of  $Y$  in which  $u$  satisfies (A). Define  $Y_{u,B}$ ,  $Y_{u,C}$  and  $Y_{u,D}$  similarly. We will show that the size of each of  $Y_{u,A}$ ,  $Y_{u,B}$ ,  $Y_{u,C}$  and  $Y_{u,D}$  is at most  $(\epsilon/(16n))|Y|$ . Equation (7) follows since

$$|Y_0| \leq \sum_{u \in V(G)} |Y_{u,A}| + |Y_{u,B}| + |Y_{u,C}| + |Y_{u,D}|.$$

First, let's show that  $|Y_{u,A}| \leq (\epsilon/(16n))|Y|$ . Consider the set of colourings in  $Y$  in which all neighbours of vertices in  $R(u)$  have colours from  $F_L$  and let  $\psi$  be the number of induced colourings on vertices other than the vertices of  $R(u)$ . If  $\psi = 0$  then  $|Y_{u,A}| = 0$ , so the claim is trivial. Otherwise,  $|Y_{u,A}| \leq \psi(v_R^b - \nu(b, v_R))$  which is at most  $\psi v_R^b \exp(-b/(2v_R))$  by Lemma 3. On the other hand,  $|Y| \geq \psi v_R^b$ , so the claim follows from Equation (10). The proof that  $|Y_{u,C}|$  is sufficiently small is similar.

Next, let's show that  $|Y_{u,B}| \leq (\epsilon/(16n))|Y|$ . Consider the set of colourings in  $Y$  in which  $R(u)$  is coloured with a subset of  $F_R$  and let  $\psi$  be the number of induced colourings on all vertices except those in  $L(u)$  and  $A_{i,u}$  and  $A'_{i,u}$  (for  $i \in [1, \dots, j+k]$ ). If  $\psi = 0$  then  $|Y_{u,B}| = 0$ , so the claim is trivial. Otherwise,  $|Y_{u,B}| \leq \psi(v_L^a - \nu(a, v_L))v_R^q v_L^q$  which is at most  $\psi v_L^a \exp(-a/(2v_L))v_R^q v_L^q$  by Lemma 3. On the other hand,  $|Y| \geq \psi v_L^a f_R^q v_L^q$ , so the claim follows from Equation (9). The proof that  $|Y_{u,D}|$  is sufficiently small is similar.

We will now work on establishing Equation (5). First consider  $i \in [1, \dots, k]$ . Let  $Y_{u,i}$  be the set of colourings in  $Y_i$  for which  $u \in V_L(G)$  is the first vertex in  $V_L(G)$  with  $R(u)$  coloured  $S_i$ . Let  $\Gamma$  be the set of induced colourings on  $B_{i,u}$ . Note that  $\Gamma$  is the set of fixed  $H_{S_i}$ -colourings of  $G_i = f_i(G, \epsilon/2)$ . Also, each colouring in  $\Gamma$  comes up  $\psi$  times in  $Y_{u,i}$  for some  $\psi$ . (In particular,  $\psi$  is the number of colourings of vertices other than  $B_{i,u}$  which are induced by colourings in  $Y_{u,i}$ .) For colouring  $y \in Y_{u,i}$  we

will let  $g(G, \epsilon, y) = g_i(G_i, \epsilon/2, y')$  where  $y'$  is the induced colouring on  $B_{i,u}$ . Then for every independent set  $x$  in the set  $\mathcal{I}(G)$  of independent sets of  $G$ ,

$$|\{y \in Y_{u,i} \mid g(G, \epsilon, y) = x\}| = \psi |\{y' \in \Gamma \mid g_i(G_i, \epsilon/2, y') = x\}|. \quad (11)$$

Since  $(f_i, g_i)$  is an SP-reduction, Equation (1) gives

$$e^{-\epsilon/2} \frac{|\Gamma|}{|\mathcal{I}(G)|} \leq |\{y' \in \Gamma \mid g_i(G_i, \epsilon/2, y') = x\}| \leq e^{\epsilon/2} \frac{|\Gamma|}{|\mathcal{I}(G)|} \quad (12)$$

and Equation (5) follows for  $Y_{u,i}$  from Equations (11) and (12) since  $|Y_{u,i}| = \psi |\Gamma|$ . Colourings in  $Y_{k+1}, \dots, Y_{k+j}$  are handled similarly except that we look at induced colourings of  $A_{i,u}$  rather than  $B_{i,u}$ .

It remains to satisfy Equation (5) for  $i = k+j+1$ . Note that any colouring  $y$  in  $Y_{k+j+1}$  points out an independent set of  $G$ . A vertex  $u \in V_L(G)$  is in the independent set if  $R(u)$  is coloured  $V_R(H)$ . A vertex  $v \in V_R(G)$  is in the independent set if  $L(v)$  is coloured  $V_L(H)$ . We will define  $g(G, \epsilon, y)$  to be this independent set. Let us focus attention on a given independent set containing  $w_L$  vertices in  $V_L(G)$  and  $w_R$  vertices in  $V_R(G)$ . We will now calculate how many colourings in  $Y_{k+j+1}$  correspond to this independent set.

For any bipartite graph  $G'$  with vertex partition  $(V_L(G'), V_R(G'))$ , let  $\phi_H(G')$  denote the number of fixed  $H$ -colourings of  $G'$ . Then the number of times that this independent set comes up as a colouring in  $Y_{k+j+1}$  is the product of the following two quantities.

$$\begin{aligned} & \left( \nu(b, v_R) f_L^a \prod_{i=1}^{k+j} \phi_H(A_{i,u}) \phi_H(A'_{i,u}) f_L^{a_i+b_i} v_R^{a_i+b_i} \right)^{w_L+v_R-w_R}, \\ & \left( f_R^b \nu(a, v_L) \prod_{i=1}^{k+j} \phi_H(B_{i,u}) \phi_H(B'_{i,u}) v_L^{a_i+b_i} f_R^{a_i+b_i} \right)^{v_L-w_L+w_R}. \end{aligned}$$

Now note that  $\phi_H(A_{i,u}) = \phi_H(B_{i,u})$  and  $\phi_H(A'_{i,u}) = \phi_H(B'_{i,u})$ . So if we let

$$Z = \left( \prod_{i=1}^{k+j} \phi_H(A_{i,u}) \phi_H(A'_{i,u}) \right)^{v_L+v_R} (f_L^a \nu(b, v_R) f_L^q v_R^q)^{v_R} (f_R^b \nu(a, v_L) v_L^q f_R^q)^{v_L},$$

the contribution of the independent set becomes

$$Z(\nu(b, v_R) f_L^a f_L^q v_R^q)^{w_L - w_R} (f_R^b \nu(a, v_L) v_L^q f_R^q)^{w_R - w_L},$$

which is

$$Z \left( \frac{\nu(b, v_R) v_L^a}{v_R^b \nu(a, v_L)} \right)^{w_L - w_R} \left( \left( \frac{v_R}{f_R} \right)^{b+q} \left( \frac{f_L}{v_L} \right)^{a+q} \right)^{w_L - w_R}.$$

To get Equation (8) we will show that  $a$  and  $b$  can be chosen so that

$$e^{-\epsilon/(8n)} \leq \left( \frac{\nu(b, v_R) v_L^a}{v_R^b \nu(a, v_L)} \right) \leq e^{\epsilon/(8n)}, \quad (13)$$

and

$$e^{-\epsilon/(8n)} \leq \left( \frac{v_R}{f_R} \right)^{b+q} \left( \frac{f_L}{v_L} \right)^{a+q} \leq e^{\epsilon/(8n)}. \quad (14)$$

This guarantees that the contribution of this independent set is in the range  $[e^{-\epsilon/4} Z, e^{\epsilon/4} Z]$ , and Equation (8) follows for  $Y_{k+j+1}$ . To establish Equation (13), use Lemma 3 to observe that

$$\left( \frac{\nu(b, v_R) v_L^a}{v_R^b \nu(a, v_L)} \right) \leq \frac{1}{1 - \exp(-a/(2v_L))}.$$

Since Equation (9) gives  $1 - \exp(-a/(2v_L)) \geq 1 - \epsilon/(16n) \geq e^{-\epsilon/(8n)}$ , the right-hand inequality of (13) follows. The left-hand inequality is similar.

We will now show how to choose the values of  $a$  and  $b$  to satisfy Equation (14). If  $v_R/f_R = v_L/f_L$  then simply choose  $a = b$  and make them large enough to satisfy Equation (9) and Equation (10). Suppose that  $v_R/f_R < v_L/f_L$ . Then use Lemma 4 with  $c_1 = v_R/f_R$ ,  $c_2 = v_L/f_L$ ,  $\delta = \epsilon/(8n)$ , and

$$a_0 = 2v_L [q \ln(v_R/f_R) + \ln(16n/\epsilon)] + 2v_R [q \ln(v_L/f_L) + \ln(16n/\epsilon)].$$

The lemma gives values of  $a$  and  $b$  which are in  $O((a_0 + q)/\delta)$ , which is not too large. Thus, our reduction is sampling-preserving. Note that the reduction can be done in polynomial time — the calculation of  $a$  and  $b$  does not involve computing  $Z$ . The case where  $v_L/f_L < v_R/f_R$  is similar.  $\square$

## 6 The proof of Theorem 2

We start with some definitions. First, for every graph  $H$ , we will define a loop-free bipartite graph  $B[H]$  (this construction was used in [10]). Let the vertices of  $H$  be  $v_1, \dots, v_h$ . The vertex set of  $B[H]$  is  $\{x_1, \dots, x_h\} \cup \{y_1, \dots, y_h\}$ . The edge set of  $B[H]$  is

$$\{(x_i, y_j) \mid (v_i, v_j) \in E(H)\}.$$

Thus, a loop  $(v_i, v_i)$  in  $H$  corresponds to the edge  $(x_i, y_i)$  in  $B[H]$  and a non-loop  $(v_i, v_j)$  in  $H$  (for which  $i \neq j$ ) corresponds to two edges  $(x_i, y_j)$  and  $(y_i, x_j)$  in  $B[H]$ . For every edge  $(v_i, v_j)$  of  $H$ , let

$$V_L(H_{i,j}) = \{x_\ell \mid (v_\ell, v_j) \in E(H)\}$$

and

$$V_R(H_{i,j}) = \{y_\ell \mid (v_i, v_\ell) \in E(H)\}$$

and let  $H_{i,j}$  be the subgraph of  $B[H]$  induced by vertex set  $V_L(H_{i,j}) \cup V_R(H_{i,j})$ . Note that  $x_i \in V_L(H_{i,j})$  and  $y_j \in V_R(H_{i,j})$  and  $x_i$  is adjacent to all of  $V_R(H_{i,j})$  in  $H_{i,j}$  and  $y_j$  is adjacent to all of  $V_L(H_{i,j})$ . Thus,  $H_{i,j}$  is connected and full. Let  $\Delta_1(H)$  be the degree of  $H$ . That is,

$$\Delta_1(H) = \max\{\deg(v) \mid v \in V(H)\}.$$

Similarly, let  $\Delta_2(H)$  be the maximum degree amongst neighbours of vertices with degree  $\Delta_1(H)$ :

$$\Delta_2(H) = \max\{\deg(v) \mid \text{for some } u \in V(H) \text{ with } \deg(u) = \Delta_1(H), (u, v) \in E(H)\}.$$

Let

$$R(H) = \{(v_i, v_j) \mid ((v_i, v_j) \in E(H) \text{ and } \deg(v_i) = \Delta_1(H) \text{ and } \deg(v_j) = \Delta_2(H))\}.$$

We will start with the following lemma.

**Lemma 6** *Let  $H$  be any fixed graph with no trivial components. Then  $R(H)$  is non-empty and  $\Delta_1(H) > 1$  and  $\Delta_2(H) > 1$ . Also, for all  $(v_i, v_j) \in R(H)$ ,  $H_{i,j}$  is connected, loop-free, bipartite, full and nontrivial.*

*Proof.* Since  $H$  has no trivial components,  $R(H)$  is non-empty and  $\Delta_1(H) > 1$  and  $\Delta_2(H) > 1$ . Suppose  $(v_i, v_j) \in R(H)$ . Recall that  $H_{i,j}$  is connected, loop-free, bipartite and full. Suppose for contradiction that  $H_{i,j}$  is a complete bipartite graph (so vertices in  $V_L(H_{i,j})$  have degree  $\Delta_1(H)$  in  $H_{i,j}$  and vertices in  $V_R(H_{i,j})$  have degree  $\Delta_2(H)$  in  $H_{i,j}$ ).

This assumption guarantees that  $H_{i,j}$  is a connected component of  $B[H]$ :  $B[H]$  cannot have an edge with exactly one endpoint in  $V_L(H_{i,j})$  — the endpoint would then have degree exceeding  $\Delta_1(H)$  in  $B[H]$ , which is a contradiction; similarly,  $B[H]$  cannot have an edge with exactly one endpoint in  $V_R(H_{i,j})$ .

Thus, for any  $x_\ell \in V_L(H_{i,j})$ ,

$$\{v_r \mid (v_\ell, v_r) \in E(H)\} = \{v_r \mid y_r \in V_R(H_{i,j})\}. \quad (15)$$

Similarly, for any  $y_\ell \in V_R(H_{i,j})$ ,

$$\{v_r \mid (v_\ell, v_r) \in E(H)\} = \{v_r \mid x_r \in V_L(H_{i,j})\}. \quad (16)$$

Now if  $H$  has a vertex  $v_\ell$  such that  $(v_i, v_\ell) \in E(H)$  and  $(v_j, v_\ell) \in E(H)$  then  $x_\ell \in V_L(H_{i,j})$  and  $y_\ell \in V_R(H_{i,j})$  so Equations (15) and (16) imply that

$$\{v_r \mid y_r \in V_R(H_{i,j})\} = \{v_r \mid x_r \in V_L(H_{i,j})\}.$$

Thus,  $H_{i,j}$  corresponds to a component of  $H$  and that component is a looped clique, which contradicts the fact that  $H$  has no trivial component.

On the other hand, if there is no  $v_\ell$  with  $(v_i, v_\ell) \in E(H)$  and  $(v_j, v_\ell) \in E(H)$  then  $H_{i,j}$  corresponds to a connected component of  $H$  which is a complete bipartite graph, again giving a contradiction.  $\square$

We can now prove the main lemma.

**Lemma 7** *Suppose that  $H$  is a fixed graph with no trivial components. Then  $\text{SAMPLEBIS} \leq_{\text{SP}} \text{SAMPLEBH-COL}$ .*

*Proof.* Let  $(G, \epsilon)$  be an input to  $\text{SAMPLEBIS}$ . For each  $(v_i, v_j) \in R(H)$ , Lemma 6 and Lemma 5 guarantee that there is a sampling-preserving reduction  $(f_{i,j}, g_{i,j})$  from  $\text{SAMPLEBIS}$  to  $\text{SAMPLEFIXEDH}_{i,j}\text{-COL}$ . Let  $G_{i,j} = f_{i,j}(G, \epsilon/2)$ . Let  $f(G, \epsilon)$  be the graph which is constructed as follows. See Figure 3. Let  $q = \sum_{(v_i, v_j) \in R(H)} |V_L(G_{i,j})| + |V_R(G_{i,j})|$ . Let

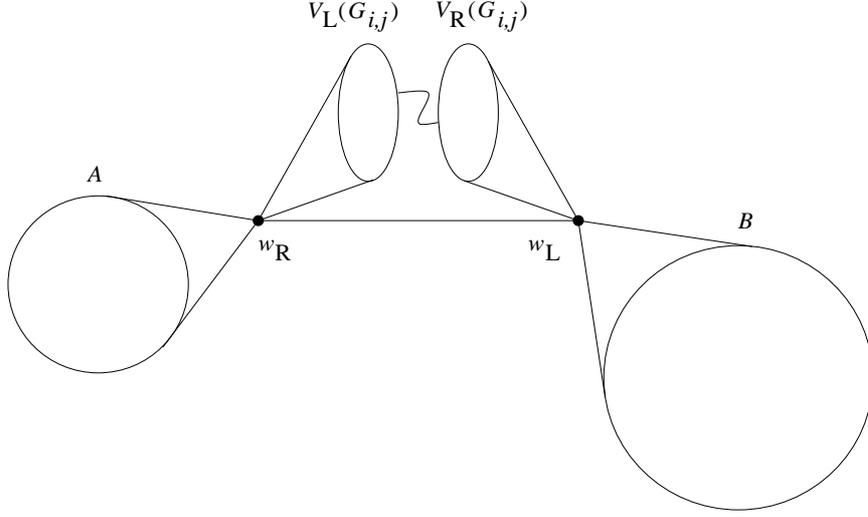


Figure 3: The construction of  $f(G, \epsilon)$  in the proof of Lemma 7.

$$V_L(f(G, \epsilon)) = A \cup \{w_L\} \cup \bigcup_{(v_i, v_j) \in R(H)} V_L(G_{i,j}),$$

and

$$V_R(f(G, \epsilon)) = B \cup \{w_R\} \cup \bigcup_{(v_i, v_j) \in R(H)} V_R(G_{i,j}),$$

where  $A$  and  $B$  are sets of vertices with

$$|A| = \left\lceil \frac{q \ln(|V(H)|) + \ln(8|E(H)|/\epsilon)}{\ln(\Delta_2(H)/(\Delta_2(H) - 1))} \right\rceil$$

and

$$|B| = \left\lceil \frac{(q + |A| + 1) \ln(|V(H)|) + \ln(8|V(H)|/\epsilon)}{\ln(\Delta_1(H)/(\Delta_1(H) - 1))} \right\rceil.$$

Note that there is no division by zero, since  $\Delta_1(H)$  and  $\Delta_2(H)$  are bigger than one (by Lemma 6). In addition to the edges in the graphs  $G_{i,j}$ , we add edge  $(w_L, w_R)$  and  $w_L \times B$  and  $w_R \times A$  and, for all  $(v_i, v_j) \in R(H)$ , we add edges  $w_L \times V_R(G_{i,j})$  and  $w_R \times V_L(G_{i,j})$ .

Let  $Y$  be the set of  $H$ -colourings of  $f(G, \epsilon)$ .  $Y_0$  will be the set of colourings in  $Y$  in which  $(w_L, w_R)$  is not coloured with an edge  $(v_i, v_j)$  from  $R(H)$ . We

will now establish Equation (7). For every  $v \in V(H)$  with  $\deg(v) < \Delta_1(H)$  let  $Y_0(v)$  be the set of colourings in  $Y$  in which  $w_L$  is coloured  $v$ . Now

$$|Y_0(v)| \leq (\Delta_1(H) - 1)^{|B|} |V(H)|^{q+|A|+1}.$$

Now consider any  $(v_i, v_j) \in R(H)$ . There are at least  $\Delta_2(H)^{|A|} \Delta_1(H)^{|B|}$  colourings of  $f(G, \epsilon)$  with  $(w_L, w_R)$  coloured  $(v_i, v_j)$  (for example, the colourings in which all of the vertices of the graphs  $G_{i,j}$  are coloured with either  $v_i$  or  $v_j$ ). Thus,  $|Y| \geq \Delta_2(H)^{|A|} \Delta_1(H)^{|B|} \geq \Delta_1(H)^{|B|}$ . We conclude that

$$|Y_0(v)| \leq (\epsilon/(8|V(H)|))|Y|. \quad (17)$$

Now consider any edge  $(v_i, v_k) \in E(H)$  such that  $\deg(v_i) = \Delta_1(H)$  but  $\deg(v_k) < \Delta_2(H)$ . Let  $Y_0(v_i, v_k)$  be the set of colourings in  $Y$  in which  $(w_L, w_R)$  is coloured  $(v_i, v_k)$ . Now

$$|Y_0(v_i, v_k)| \leq \Delta_1(H)^{|B|} (\Delta_2(H) - 1)^{|A|} |V(H)|^q.$$

Also, from before  $|Y| \geq \Delta_2(H)^{|A|} \Delta_1(H)^{|B|}$  so

$$|Y_0(v_i, v_k)| \leq (\epsilon/(8|E(H)|))|Y|. \quad (18)$$

Equation (17) and (18) imply Equation (7) since  $|Y_0| \leq \sum_{v \in V(H)} Y_0(v) + \sum_{(v_i, v_k)} |Y_0(v_i, v_k)|$ .

For an edge  $(v_i, v_j) \in R(H)$ , let  $Y_{i,j}$  be the set of colourings of  $f(G, \epsilon)$  with  $(w_L, w_R)$  coloured  $(v_i, v_j)$ . Let  $\Gamma$  be the set of induced colourings on  $G_{i,j}$ . Note that  $\Gamma$  is the set of fixed  $H_{i,j}$ -colourings of  $G_{i,j}$ . Also, each colouring in  $\Gamma$  comes up  $\psi$  times in  $Y_{i,j}$  where  $\psi$  is the number of induced colourings on the vertices other than  $G_{i,j}$ . For a colouring  $y \in Y_{i,j}$  we will set  $g(G, \epsilon, y) = g_{i,j}(G_{i,j}, \epsilon/2, y')$  where  $y'$  is the induced colouring on  $G_{i,j}$ . Then Equation (5) follows from the fact that  $(f_{i,j}, g_{i,j})$  is an SP-reduction.  $\square$

Theorem 2 follows from Lemma 1 and Lemma 7 and from Lemma 8 below. Recall the following definitions. A *randomised approximation scheme* (RAS) for a counting problem  $F$  is a randomised algorithm that takes input  $\sigma$  and accuracy parameter  $\epsilon \in (0, 1)$  and produces as output an integer random variable  $Y$  satisfying the condition  $\Pr(e^{-\epsilon} F(\sigma) \leq Y \leq e^\epsilon F(\sigma)) \geq 3/4$ . It is a “fully polynomial” randomised approximation scheme (FPRAS) if it runs in time  $\text{poly}(|\sigma|, \epsilon^{-1})$ . The problem  $\#\text{BIS}$  is “self-reducible” so the following lemma follows from [17].

**Lemma 8** (*JVV*) *If SAMPLEBIS has a PAUS then #BIS has an FPRAS.*

*Proof.* The lemma is a special case of Theorem 6.4 of [17]. In order to apply Theorem 6.4 directly we would need to define “self-reducible” formally and to deal with some easy (though annoying) issues:

- (i) Inputs to #BIS may be disconnected but inputs to SAMPLEBIS may not.
- (ii) In order to apply Theorem 6.4 we technically need a *fully* polynomial almost uniform sampler (FPAUS) for SAMPLEBIS. This can be obtained from a PAUS as [17] explains.

Rather than dealing with these issues, we prefer to simply provide a proof for the lemma. The details given here are from the proof of Proposition 3.4 of [16]. Technically, Jerrum’s proof from [16] is about counting *matchings* but the few changes that are needed to yield our lemma are completely routine.

Let  $(G, \epsilon)$  be an input to #BIS. Suppose that  $G$  has components  $G_1, \dots, G_k$ . For each  $i$ , let the two parts of the vertex set be  $V_L(G_i)$  and  $V_R(G_i)$  and let the sizes of these parts be  $\ell_i$  and  $r_i$ , respectively. Let  $N_i = \ell_i r_i$  and let  $E(G_i) = \{e_i(1), \dots, e_i(m_i)\}$ . Denote the non-edges of  $G_i$  by  $\{e_i(m_i + 1), \dots, e_i(N_i)\}$ . For  $j \in \{1, \dots, N_i\}$ , let  $G_i(j)$  be the graph  $(V(G_i), \{e_i(1), \dots, e_i(j)\})$ . For any graph  $G'$ , let  $\mathcal{I}(G')$  denote the set of independent sets of  $G'$ . Let

$$\rho_i(j) = \frac{|\mathcal{I}(G_i(j+1))|}{|\mathcal{I}(G_i(j))|}.$$

Note that

$$|\mathcal{I}(G_i)| = (\rho_i(m_i)\rho_i(m_i + 1) \cdots \rho_i(N_i - 1))^{-1} |\mathcal{I}(G_i(N_i))|.$$

Also, the number of independent sets of the complete bipartite graph  $G_i(N_i)$  is  $2^{\ell_i} + 2^{r_i} - 1$ , so

$$|\mathcal{I}(G_i)| = (2^{\ell_i} + 2^{r_i} - 1) \prod_{j=m_i}^{N_i-1} \rho_i(j)^{-1}. \quad (19)$$

Furthermore,

$$|\mathcal{I}(G)| = \prod_{i=1}^k |\mathcal{I}(G_i)| = \prod_{i=1}^k (2^{\ell_i} + 2^{r_i} - 1) \prod_{j=m_i}^{N_i-1} \rho_i(j)^{-1}. \quad (20)$$

Now let  $z = \sum_{i=1}^k (N_i - m_i)$ . In order to estimate  $|\mathcal{I}(G)|$ , we need to estimate the  $z$  ratios  $\rho_i(j)$ .

For each ratio  $\rho_i(j)$  we can make some observations.

- (i)  $\rho_i(j) \leq 1$ , since  $\mathcal{I}(G_i(j+1)) \subseteq \mathcal{I}(G_i(j))$
- (ii)  $\rho_i(j) \geq 1/2$ , since  $\mathcal{I}(G_i(j)) \setminus \mathcal{I}(G_i(j+1))$  can be mapped injectively into  $\mathcal{I}(G_i(j+1))$  by removing the lexicographically-least endpoint of  $e_i(j+1)$ .
- (iii) Let  $\mathcal{A}$  be a PAUS for SAMPLEBIS. For  $i \in [1, \dots, k]$  and  $j \in [m_i, \dots, N_i - 1]$ , let  $Z_i(j)$  be the indicator variable for the event that, when we run  $\mathcal{A}$  with input  $G_i(j)$  and accuracy parameter  $\delta$ , the output is an independent set of  $G_i(j+1)$ . Note that  $\rho_i(j) - \delta \leq E[Z_i(j)] \leq \rho_i(j) + \delta$ . This follows immediately from the definition of PAUS, but it is important to note that the input to  $\mathcal{A}$ ,  $G_i(j)$ , is connected (since all inputs to SAMPLEBIS must be connected).

Let  $\overline{Z_i(j)}$  be the result obtained by calling  $\mathcal{A}$   $\lceil 74\epsilon^{-2}z \rceil$  times with input  $G_i(j)$  and accuracy parameter  $\delta = \epsilon/(6z)$  and averaging the value of  $Z_i(j)$  which occurs each time. Jerrum shows in his proof that with probability at least  $3/4$ ,

$$e^{-\epsilon} \prod_{i=1}^k \prod_{j=m_i}^{N_i-1} \rho_i(j) \leq \prod_{i=1}^k \prod_{j=m_i}^{N_i-1} \overline{Z_i(j)} \leq e^{\epsilon} \prod_{i=1}^k \prod_{j=m_i}^{N_i-1} \rho_i(j).$$

Thus, the quantity

$$\prod_{i=1}^k (2^{\ell_i} + 2^{r_i} - 1) \prod_{j=m_i}^{N_i-1} \overline{Z_i(j)}^{-1}$$

is a sufficiently accurate estimate of  $|\mathcal{I}(G)|$ .

For each of the  $z$  pairs  $(i, j)$ ,  $O(\epsilon^{-2}z)$  samples were needed, each of which is produced in time  $\text{poly}(|G|, z/\epsilon)$ . Since  $z \leq |V(G)|^2$ , the total running time is  $\text{poly}(|G|, \epsilon^{-1})$  and we have an FPRAS.  $\square$

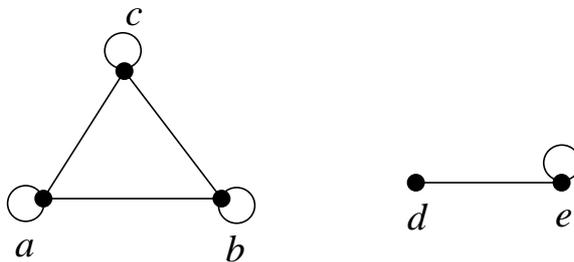


Figure 4: An  $H$  with a nontrivial component for which  $\text{SAMPLEH-COL}$  has a PAUS.

## 7 The presence of trivial components

Theorem 2 shows that sampling  $H$ -colourings is difficult if every component of  $H$  is nontrivial. Recall from [10] that exactly counting  $H$ -colourings is  $\#P$ -complete if  $H$  has even one nontrivial component. Thus, it might appear that Theorem 2 can be improved. In this section, we show that the existence of a single nontrivial component is not enough to make sampling difficult. In particular, we give an example of a graph  $H$  with a nontrivial component, for which  $\text{SAMPLEH-COL}$  has a PAUS. Specifically, let  $H$  be the graph depicted in Figure 4.

**Observation 9** *Suppose that  $H$  is the graph depicted in Figure 4.  $\text{SAMPLEH-COL}$  has a PAUS.*

*Proof.* Here is a PAUS for  $\text{SAMPLEH-COL}$ . The input is an instance  $(G, \epsilon)$  where  $G$  has  $n$  vertices and, without loss of generality<sup>3</sup>, is connected. If  $\epsilon < 2^n / (2^n + 3^n)$  then the algorithm simply runs for  $5^n$  steps, constructs all of the  $H$ -colourings of  $G$  (and counts them) and selects one uniformly at random. Note that the running time is at most  $\text{poly}(1/\epsilon)$  in this case. Otherwise, the algorithm chooses  $i$  uniformly at random from  $1, \dots, 3^n + 2^n$ . If  $i \leq 3^n$ , then the algorithm outputs the  $i$ 'th colouring from the  $3^n$  colourings with colours “ $a$ ”, “ $b$ ”, and “ $c$ ”. Otherwise, let  $C$  be the  $(i - 3^n)$ th of the  $2^n$  (proper and

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<sup>3</sup>We can assume that the input is a connected graph without losing generality because we can obtain an  $H$ -colouring of a  $k$ -component graph  $G$  by independently calling our PAUS for each component, specifying accuracy parameter  $\epsilon/k$ . The final variation distance (between the output distribution and the uniform distribution on  $H$ -colourings of  $G$ ) is at most  $\epsilon$ .

improper) colourings with colours “ $d$ ” and “ $e$ ”. If  $C$  is a legal  $H$ -colouring of  $G$ , then the algorithm outputs it. Otherwise, it outputs the error symbol  $\perp$ . The variation distance between the output distribution of the algorithm and the uniform distribution on  $H$ -colourings of  $G$  is at most the probability that the algorithm outputs  $\perp$ , which is at most  $2^n/(2^n + 3^n) \leq \epsilon$ .  $\square$

## 8 Sampling and Counting

Let  $\#BH\text{-COL}$  be defined as follows.

*Name.*  $\#BH\text{-COL}$ .

*Instance.* A loop-free connected bipartite graph  $G$ .

*Output.* The number of  $H$ -colourings of  $G$ .

For certain graphs  $H$ , the problem  $\#BH\text{-COL}$  can be expressed as the counting problem associated with a “self-reducible  $p$ -relation”. For such an  $H$ , Theorem 6.3 of Jerrum, Valiant and Vazirani’s paper [17] guarantees that if there is an FPRAS for  $\#BH\text{-COL}$  then there is a PAUS for  $\text{SAMPLE}BH\text{-COL}$ . If  $H$  has no trivial components, this in turn guarantees (by Theorem 2) an FPRAS for  $\#BIS$ . Dyer and Greenhill [8] have given a more general framework in which these ideas work: If, for a given graph  $H$ , the problem  $\#BH\text{-COL}$  is “self-partitionable” then an FPRAS for  $\#BH\text{-COL}$  can be turned into a PAUS for  $\text{SAMPLE}BH\text{-COL}$ . It is not clear for which graphs  $H$  these ideas can be applied, and this is an interesting open question.

A related problem (which is also open) is to determine for which graphs  $H$  an FPRAS for counting  $H$ -colourings can be turned into a PAUS for  $\text{SAMPLE}H\text{-COL}$ . Dyer, Goldberg and Jerrum [7] have shown that for every fixed  $H$  a PAUS for  $\text{SAMPLE}H\text{-COL}$  can be turned into an FPRAS for counting  $H$ -colourings.

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