



Rejection sampling of bipartite graphs with given degree sequence

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Abstract. Let $A = (a_1, a_2, \dots, a_n)$ be a degree sequence of a simple bipartite graph. We present an algorithm that takes A as input, and outputs a simple bipartite realization of A , without stalling. The running time of the algorithm is $\Theta(n_1 n_2)$, where n_i is the number of vertices in the part i of the bipartite graph. Then we couple the generation algorithm with a rejection sampling scheme to generate a simple realization of A uniformly at random. The best algorithm we know is the implicit one due to Bayati, Kim and Saberi (2010) that has a running time of $\mathcal{O}(m a_{\max})$, where $m = \frac{1}{2} \sum_{i=1}^n a_i$ and a_{\max} is the maximum of the degrees, but does not sample uniformly. Similarly, the algorithm presented by Chen et al. (2005) does not sample uniformly, but nearly uniformly. The realization of A output by our algorithm may be a start point for the edge-swapping Markov Chains pioneered by Brualdi (1980) and Kannan et al. (1999).

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1 Introduction

A graph $G(V(G), E(G))$ is said to be *bipartite* if its vertex set $V(G)$ can be partitioned into two different sets $V_1(G)$ and $V_2(G)$ with $V(G) = V_1(G) \cup V_2(G)$ such that $uv \in E$ if $u \in V_1$ and $v \in V_2$. The graphs considered can have possible parallel edges and loops unless otherwise stated. The *Degree Sequence Problem* is to find some or all graphs with a given degree sequence [30, 34]. More detailed analysis of the Degree Sequence Problem and its relevance can be found in [29]. It is much researched upon for its relevance in network modelling in Ecology, Social Sciences, chemical compounds and biochemical networks in the cell. Especially, ecological occurrence matrices, such as the Darwin finches tables, are $(0, 1)$ matrices whose rows are indexed by species of animals and columns are islands, and the (i, j) entry is 1 if animal i lives in island j , and is 0 otherwise. Moreover the row sums and columns sums are fixed by field observation of these islands. These occurrence matrices are thus bipartite graphs G with a fixed degree sequence in which $V_1(G)$ is the set of animals and $V_2(G)$ is the set of islands. Researchers in Ecology [8, 9, 15, 31] are highly interested in sampling easily and uniformly ecological occurrence tables, so that by using Monte Carlo methods, they can approximate test statistics to prove or disprove some null hypothesis about competitions amongst animals. Several algorithms are known to sample random realizations of degree sequences, and each one of them has its strengths and limitations. Most of these use Monte Carlo Markov chain methods based on edge-swapping [6, 9, 10, 11, 12, 13, 18, 22, 21, 24]. Since to start a Markov chain still requires to have a realisation of the degree sequence A , many algorithms are proposed that generate such a realisation [1, 3, 5, 2, 36]. Most of these algorithms are based on random matching methods. In particular, algorithms proposed in [1, 3, 8] are based on inserting edges sequentially according to some probability scheme. The basic ideas of the algorithm presented in the present paper can be seen as implementing a "dual sequential method", as it inserts sequentially vertices instead of edges.

In the theory of the Tutte polynomial, there are two operations, deletion and contraction, that are dual to each other, see [7] for more details on this topic. Let G be a graph having n vertices and m edges. In G , the operation of deleting an edge $e = (v_i, v_j)$ means removing the edge e and the graph thus obtained, denoted by $G \setminus e$, is a graph on n vertices and $m - 1$ edges where both the degrees of vertices v_i and v_j decrease by 1. The operation of contracting the graph G by $e = (v_i, v_j)$ consists of deleting the edge e and identifying the vertices v_i and v_j . The graph thus obtained, denoted by G/e , is a graph on $n - 1$ vertices and $m - 1$ edges where the new vertex obtained by identifying

v_i and v_j has degree $\alpha_i + \alpha_j - 2$. Deletion is said to be the dual of contraction as the incidence matrix of $G \setminus e$ is orthogonal to the incidence matrix of G^* / e , where G^* is the dual of G if G is planar.

If A is a degree sequence having n entries, it can easily be shown that random matching methods used in [1, 2, 3, 5, 36] are equivalent to starting from a known realization G of A , delete all the edges one by one, and keeping track of the degrees of vertices after each deletion, until one reaches the empty graph having n vertices. Then, reconstructing a random realization of A consists of taking the reverse of the deletion. That is, starting from the empty graph on n vertices, re-insert edges one by one by choosing which edge to insert according to the degrees of the vertices and some probability scheme depending on the stage of the algorithm, and subject to not getting double edges if one would like to get simple graphs or not linking two vertices on the same part if one wants to get bipartite graphs. The algorithm presented in this paper is based on the dual operation of contraction that has been slightly modified to suit our purpose. It is equivalent to starting from a known realization G of A , contract all the edges one by one, and keeping track of the vertices after each contraction, until one reaches the graph with one vertex and $\frac{1}{2} \sum_1^n \alpha_i$ loops. Then, reconstructing a random realization of A consists of reversing the process of contraction. That is, starting from a graph with one vertex and $\frac{1}{2} \sum_1^n \alpha_i$ loops, the algorithm re-inserts vertices one by one by choosing the vertices to be joined according to the degrees of the vertices and some probability that depends on the stage of the algorithm. But, to construct a bipartite realization, we force the algorithm to insert first all the vertices in $V_1(G)$ and then all the vertices in $V_2(G)$.

While algorithms that are based on reversing the deletion operation [1, 3] are easy to implement, our algorithm seems more complex as one has to satisfy not only the degree conditions on the vertices, but also some added graphical structures imposed by the contraction. But this is more of a bonus than an inconvenience, as, apart from the fact that the running time is even better, the extra structure allows an easier analysis of the algorithm. Moreover, the internal structure imposed by the contraction operation allows the algorithm to avoid most of the shortcomings of the previous algorithms. In fact, not only the algorithm never restarts, but the algorithm also allows to sample all bipartite realizations with equal probabilities, making their approximate counting much easier than by the importance sampling used in [1, 3]. Better still, this technique can be extended to construct k -partite realizations of a k -partite degree sequence A , for $k \geq 3$, where a k -partite degree sequence is defined in a natural way extending the definition of a bipartite degree sequence.

The present paper uses the following notations and terminology. Two edges e and f in $E(G)$ are said to be *multiple* edges if they have the same end vertices (in Matroid Theory, multiple edges are said to be *parallel*). A *simple* bipartite graph is without multiple edges and contains no loops. The *degree* a_i of a vertex v_i is the number of edges incident to v_i with a loop contributing twice to the degree of v_i . The degree sequence of a graph G is formed by listing the degrees of vertices of G . If $A = (a_1, a_2, \dots, a_n)$ is a sequence of integers and G is a bipartite graph that has A as its degree sequence, we say that G is a *realization* of A , and such a sequence of integers is called a *bipartite degree sequence*. Thus entries of A can be partitioned as A_1 and A_2 , where A_i denotes the degree sequence of the part $V_i(G)$. We write V_i and $|A_i|$ to denote the set of vertices with degrees in A_i and the sum of entries in A_i respectively. In the sequel, we denote a bipartite degree sequence A as $(A_1 : A_2)$ and the pair $(A_1 : A_2)$ is called a *bipartition* of A .

Remark 1 If $A = (A_1 : A_2)$ is a bipartite degree sequence having n entries, and A_1 and A_2 have respectively n_1 and n_2 entries, then the following are true.

1. $n_1 + n_2 = n$
2. $|A_1| = |A_2|$.
3. The maximal entry of A_1 is less or equal to n_2 and vice versa.

Conversely, any partition of entries of A into two sets B_1 and B_2 satisfying Observation 1 is a bipartition of A .

In the sequel, we make use of Rejection Sampling to sample all realizations of the degree sequence with equal probability. Indeed, let $\mathcal{S} = S_1, \dots, S_r$ be a set of structures, where S_i is obtained with probability $\pi(S_i)$ such that $\sum_i \pi(S_i) = 1$. That is, the set of $\pi(S_i)$ is a probability distribution function. Let $\min(\pi)$ be the minimal probability amongst all $\pi(S_i)$. The Rejection Sampling scheme consists of generating S_i , then accept it with probability $\frac{\min(\pi)}{\pi(S_i)}$ or reject it with probability $1 - \frac{\min(\pi)}{\pi(S_i)}$. It is easy to see that every structure would then be sample with the same probability $\min(\pi)$.

This paper is organized as follows. We first define what is called a recursion chain of a degree sequence, then we present routines for constructing all bipartite realizations. The next section presents criteria and routines to generate simple bipartite realizations only. Then these basic routines are coupled with a rejection sampling routine to get a uniform distribution on the set of all simple bipartite realizations.

2 Construction all bipartite realizations of given degrees

2.1 Recursion chain of degree sequences

Let G be a graph with n vertices and m edges. Throughout we assume that the vertices and edges of G are labelled v_1, v_2, \dots, v_n . Let $A = (a_1, \dots, a_n)$ be the degree sequence of G , where a_i is the degree of the vertex v_i . Define an arithmetic operation on A , called *contraction*, as follows. For an ordered pair (a_i, a_j) of entries a_i and a_j of A with $i \neq j$, the operation of *contraction* by (a_i, a_j) means changing a_i to $a_i + a_j$ and deleting the entry a_j from A . We write $A/(i, j)$ to denote the new sequence thus obtained. We call the sequence $A/(i, j)$ the (i, j) -*minor* or simply a *minor* of A . The following example illustrates this operation for a bipartite degree sequence.

Example 1 Let $A = (4, 3, 3 : 3, 3, 2, 2)$, where $a_1 = 4, a_2 = 3, a_3 = 3$ and $a_4 = 3, a_5 = 3, a_6 = 2, a_7 = 2$. We have $A/(1, 2) = (7, 3, 3, 3, 2, 2)$ and $A/(4, 2) = (4, 3, 6, 3, 2, 2)$.

Let A be the sequence of integers. A is said to be *graphic* if there is a graph G , not necessarily bipartite, such that G has A as its degree sequence. Moreover, it is trivial to observe that a sequence of integers is graphic if and only if the sum of its entries is even.

Theorem 1 A sequence A is graphic if and only if all its minors are graphic.

Proof. Obviously, if A is graphic, then $A/(a_i, a_j)$ is graphic, as the sum of its entries is even, by definition of contraction. Now suppose that $A/(a_i, a_j)$ is graphic and G'' is a realization of $A/(a_i, a_j)$. To prove that A is also graphic, we present an algorithm, much used in the sequel, that constructs a realization of A , denoted by G , from G'' .

Algorithm AddVertex()

Step 1. To G'' add an isolated vertex labelled v_j (as in Figure 1).

Step 2 If the degree of v_j is a_j , stop, output G . Else

Step 3. Amongst the a'_i edges incident to v_i , counting loops twice, choose one edge

$e = (v_i, v_k)$ with probability $\pi(e)$ and connect e to v_j so that e becomes (v_j, v_k) . Go to Step 2.

Now, in G the degree of v_j is a_j by Step 2 of algorithm AddVertex(). Moreover, by the definition of contraction the degree of v_i is equal to $a_i + a_j$ in

G'' . Since $\text{AddVertex}()$ takes a_j edges away from v_i , the degree of v_i is a_i in G . Moreover all other vertices are left unchanged by $\text{AddVertex}()$. Thus G is a realization of A . \square

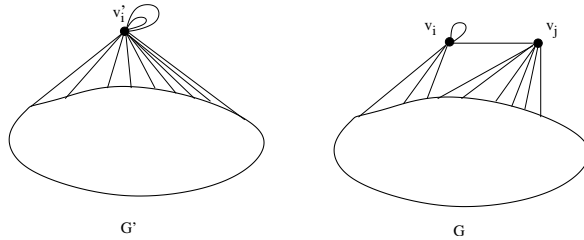


Figure 1: Construction of a graph G from its contract-minor G''

To help the intuition, observe that if G'' is a realization of $A/(a_i, a_j)$ and G is a realization of A constructed by $\text{AddVertex}()$, then G'' is obtained from G by contraction of the edge (v_i, v_j) . Now, mimicking the process of recursive contraction of matroid as used in the theory of the Tutte polynomial, we define a process of recursive contraction for a degree sequence. A *recursion chain* of a degree sequence A is a unary tree rooted at A where nodes are integer sequences and every node, except for the root, is a minor of the preceding one. The recursive procedure of contraction is carried on from the root A until a node with a single entry is reached. See Figure 2 for an illustration.

As for the Tutte polynomial, the amazing fact, which is then used to construct all the realizations of A is that the order of contraction is immaterial. Despite this basic fact, we still impose a particular order to ease many proofs in the sequel.

Notes on notations. For the sake of convenience, we denote by $A^{(i)}$ the node of a recursion chain of a degree sequence A , where i is the number of entries in the node. Thus we denote the root A by $A^{(n)}$, the next node by $A^{(n-1)}$, and so on until the last node $A^{(1)}$. Similarly, we denote by $G^{(i)}$ the realization of $A^{(i)}$. The n entries of A are labelled from 1 to n . To keep track of the vertices, we preserve the labelling of entries of A into its minors so that when a contraction by the pair (a_i, a_j) is performed, the new vertex is labelled a_i , the label a_j is deleted, and all other entries keep the labelling they have before the contraction. In this paper, we consider the recursion chain, called the *accumulating recursion chain*, constructed as follows. Let $A = (A_1 : A_2)$. We order $A = (a_1, a_2, \dots, a_n)$ as $(b_1, b_2, \dots, b_{n_1} : c_1, c_2, \dots, c_{n_2})$, where $A_1 = (b_1, b_2, \dots, b_{n_1})$ and $A_2 = (c_1, c_2, \dots, c_{n_2})$, such that $b_1 \geq b_2 \geq \dots \geq b_{n_1}$ and $c_1 \geq c_2 \geq \dots \geq c_{n_2}$ and $n_1 + n_2 = n$. Below is the pseudocode for the

recursive construction of the accumulating recursion chain of a bipartite degree sequence.

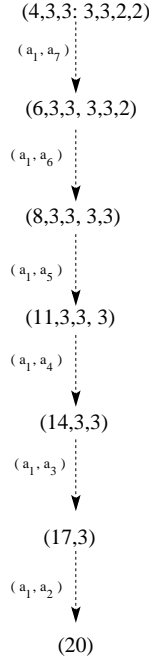


Figure 2: The recursion chain of the bipartition $(4, 3, 3 : 3, 3, 2, 2)$. Nodes of the chain are labelled from $A^{(7)} = A$ to $A^{(1)}$. Notice that we only perform contractions (v_1, v_{last}) .

Algorithm ConstructBipartiteRecursionChain()

Given a bipartite degree sequence $A = (a_1, a_2, \dots, a_n) = (b_1, b_2, \dots, b_{n_1} : c_1, c_2, \dots, c_{n_2})$ with $b_1 \geq b_2 \geq \dots \geq b_{n_1}$ and $c_1 \geq c_2 \geq \dots \geq c_{n_2}$. Let $i = n$.

Step 1 If $i = 1$, stop, return $\{A^{(1)}, A^{(2)}, \dots, A^{(n)}\}$. Else

Step 2 Let $A^{(i-1)} = A^{(i)} / (1, i)$. That is, get the $(i-1)^{\text{th}}$ recursive minor of A by contracting the $(i)^{\text{th}}$ recursive minor by its first entry and the last entry.

Step 3 Decrement i by 1 and go back to Step 1.

The accumulation recursion chain of A is denoted by $W = (A^{(1)}, A^{(2)}, \dots, A^{(n)})$.

The following algorithm generates all the bipartite realizations of A . The graph constructed is not necessarily simple. Loosely speaking, this algorithm consists of reversing the recursive process of contraction as implemented by ConstructBipartiteRecursionChain(). This algorithm starts from $G^{(1)}$ the sole

realization of $A^{(1)}$, and by calling `AddVertex()` recursively it constructs $G^{(2)}$, then $G^{(3)}$, and so on until $G^{(n)}$ that is a realization of $A^{(n)} = A$. The only conditions imposed on the choice of edges is that up to the n_1^{th} iteration, only edges (v_1, v_j) , with $j \leq n_1$, are constructed. That is, we insert vertices of V_1 . After the n_1^{th} iteration, only edges (v_k, v_j) , with $k > n_1$ and $j \leq n_1$, are constructed. That is, we insert vertices of V_2 . We call the graphs $G^{(1)}, G^{(2)}, \dots, G^{(n)}$ the *partial realizations* of A .

Algorithm ConstructBipartiteRealization()

Given $W = (A^{(1)}, A^{(2)}, \dots, A^{(n)})$, the bipartite accumulating recursion chain of A , do the following.

- Step 1.* Let $i = 1$ and build the realization of the node $A^{(1)}$, denoted by $G^{(1)}$, which is the graph consisting of one vertex and m loops, where $m = \frac{1}{2} \sum_{i=1}^n a_i$.
- Step 2.* Let $G = G^{(i)}$. If G has n vertices, stop, return G . Else,
- Step 3.* Using $G^{(i)}$ and $A^{(i+1)}$ as input, Call Algorithm `AddVertex()` to construct $G^{(i+1)}$ as a realization of $A^{(i+1)}$. If $i \leq n_1$, `AddVertex()` only concedes loops. If $i > n_1$ `Addvertex()` concedes only edges (v_1, v_j) with $1 \leq j \leq n_1$. Increment i by 1, go back to Step 2.

See Figure 3 for an illustration of Algorithm `ConstructBipartiteRealization()`.

The following definitions are needed in the sequel. In the process of contraction implemented by the accumulating recursion chain, we observe that the degrees are accumulating on a_1 . If we think of recursive contractions of a graph, this is equivalent to saying that the edges are accumulating on v_1 as v_1 seems to swallow the other vertices one by one. Hence when reversing the contraction operation in `ConstructBipartiteRealization()`, vertex v_1 plays the role of the 'mother that spawns' all the other vertices one by one and concedes some edges to them according to their degrees. Thus `AddVertex()` can attach an edge e to a new vertex v_s only if e is incident to v_1 . This observation prompts the following formal definitions. Let $A = (A_1 : A_2)$ be a bipartite degree sequence, where A_1 and A_2 have respectively n_1 and n_2 entries such that $n_1 + n_2 = n$. Up to the n_1^{th} iteration of `ConstructBipartiteRealization()`, an edge is *available* if it is a loop incident to v_1 . An edge e is *lost* otherwise. From the $(n_1 + 1)^{\text{th}}$ iteration of `ConstructBipartiteRealization()` onwards, an edge is *available* if it is incident to v_1 and a vertex v_j with $1 \leq j \leq n_1$. An edge e is *lost* otherwise. In the obvious way, we say that a vertex is *available* if it is incident to some available edge. Let V_{av} , E_{av} and E_{v_j} respectively denote the

set of all available vertices, the set of all available edges and the set of available edges that are incident to the vertex v_j , for $j \leq n_1$. An edge $e = (v_1, v_j)$ is *conceded* if $\text{AddVertex}()$ disconnects it from v_1 so that e becomes $e = (v_j, v_k)$ for some vertex $v_k \neq v_1$. We then say that v_1 (or sometimes E_{v_j} or just v_j) concedes the edge e . A vertex v_s having degree a_s is *fully inserted* if a_s edges are conceded to it. A graph G is said to be *(re)constructed* if it is an output of $\text{ConstructBipartiteRealization}()$.

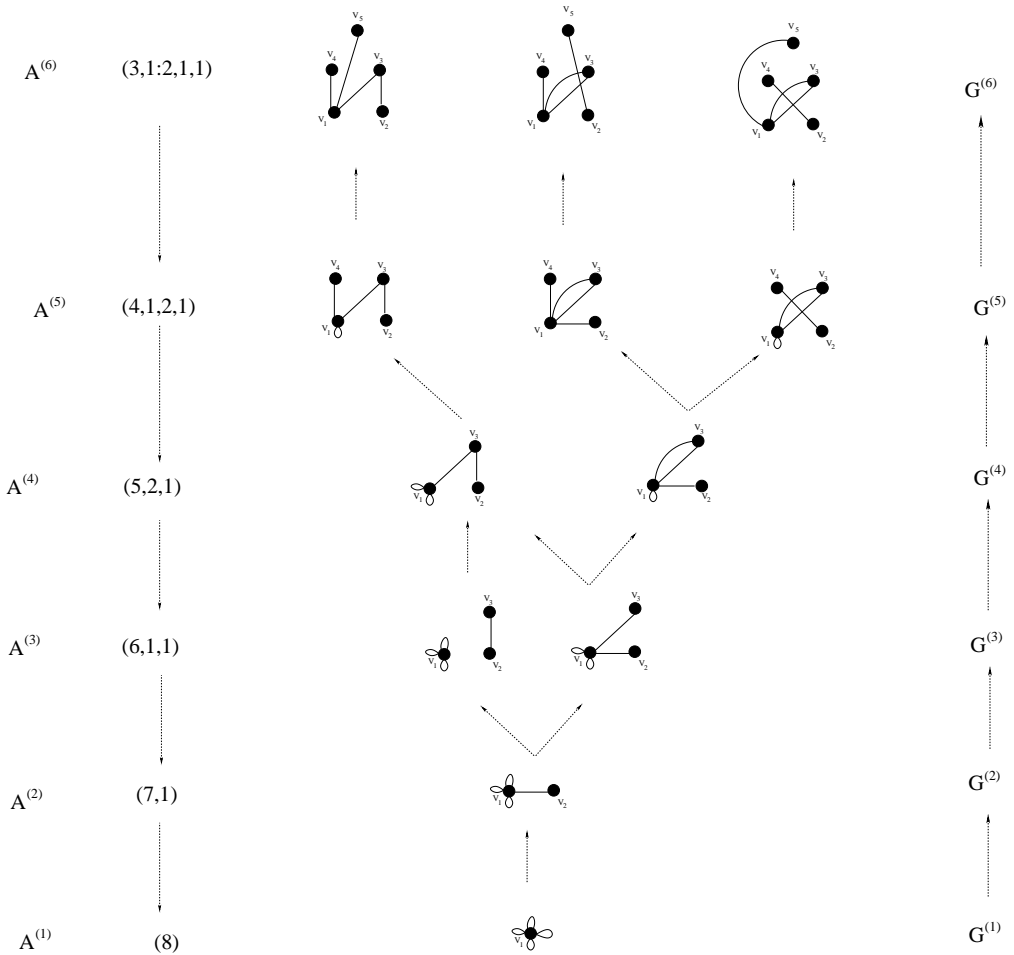


Figure 3: Random reconstruction tree of $(3, 1 : 2, 1, 1)$. Graphs drawn on the same height as the degree sequence $A^{(i)}$ corresponds to all the graphs having $A^{(i)}$ as their degree sequence. Notice that only realizations of $A^{(6)}$ are bipartite.

The next observation is an obvious consequence of the definition of the algorithm `ConstructBipartiteRealization()`. We single it out for the sake of clarity as it is used in the sequel.

Remark 2 *From the $(n_1 + 1)^{\text{th}}$ iteration of `ConstructBipartiteRealization()`, the number of available edges is equal to the number of edges left to be inserted until `ConstructBipartiteRealization()` terminates.*

It is because the number of available edges at the end of $(n_1)^{\text{th}}$ iteration is equal to half the sum of degrees $\mathbf{a}_i \in A_1$, and by the definition of the bipartite degree sequence, this number is equal to half the sum of degrees $\mathbf{a}_j \in A_2$.

Theorem 2 *Let $A = (\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n) = (A_1 : A_2)$ be a bipartite degree sequence having n entries where A_1 and A_2 respectively have n_1 and n_2 entries, such that $n_1 + n_2 = n$. Let W be the bipartite recursion chain of A . Then Algorithm `ConstructBipartiteRealization()` constructs in time linear on $m = \frac{\mathbf{a}_1 + \mathbf{a}_2 + \dots + \mathbf{a}_n}{2}$ a bipartite graph G having n vertices and m edges such that G is a realization of A . Moreover, every bipartite realization of A can be constructed in this way.*

Proof. By Algorithm `AddVertex()`, the graph $G^{(n)}$ output by Algorithm `ConstructBipartiteRealization()` is assured to be a realization of A . We need only to prove that $G^{(n)}$ is bipartite. Now, since up to the n_1^{th} iteration of `ConstructBipartiteRealization()`, the routine `AddVertex()` always chooses loops incident to v_1 , vertices inserted from the second iteration up to the n_1^{th} iteration of `ConstructBipartiteRealization()` (i.e., vertices in V_1) can never be adjacent to each other. Moreover, from the $(n_1 + 1)^{\text{th}}$ to the n^{th} iteration, `AddVertex()` never chooses an edge (v_1, v_j) with $j > n_1$. Thus all the vertices inserted from the $(n_1 + 1)^{\text{th}}$ iteration onwards (i.e., vertices in V_2) are never adjacent to each other. Thus, we only have to show that (in $G^{(n)}$), v_1 is not adjacent to any vertex inserted before the n_1^{th} iteration of `AddVertex()`. So, suppose that $G^{(n)}$ contains an edge $e = (v_1, v_j)$ with $j \leq n_1$. But, at the beginning of the $(n_1 + 1)^{\text{th}}$ iteration, the number of all edges incident to v_1 is equal to the sum of the degrees of the vertices left to insert until the end of the Algorithm. Thus one vertex v_j with $j > n_1$ is not fully inserted. This is a contradiction.

It remains to prove that any bipartite realization G of A can be constructed in this way. So, let G be a realization of A and let $e = (v_i, v_j)$, where $v_i \in V_1$ and $v_j \in V_2$, be any edge of G such that vertex v_i has degree \mathbf{a}_i and vertex v_j has degree \mathbf{a}_j . Also suppose that vertex v_i and v_j respectively were inserted at

the i^{th} and j^{th} iteration of `ConstructBipartiteRealization()`, with $i \leq n_1$ and $j > n_1$. We need to show that at the j^{th} iteration, there is a positive probability to have an edge e that is incident to v_i and e is available. Assume to the contrary, that is, at the j^{th} iteration all the edges incident to v_i must be lost. Now all the edges incident to v_i are lost before that j^{th} iteration only if at some stage of the running of `Algorithm ConstructBipartiteRealization()`, there are only the edges that are available and these are exhausted before reaching the j^{th} iteration. Thus, at the j^{th} iteration there are no more available edges. That is, there is no edge incident to v_1 . But this means that $a_{n_1+1} + a_{n_1+2} + \dots + a_{j-1} \geq m$, contradicting Observation 2.

As for the running time, `Algorithm ConstructBipartiteRealization()` calls `Algorithm AddVertex()` once for every new vertex v_k to be inserted. If v_k has degree a_k , `Algorithm AddVertex()` has to go through a_k iterations to insert the a_k edges of v_k . Hence the total number of iterations to terminate `ConstructBipartiteRealization()` is $a_1 + a_2 + \dots + a_n = 2m$. \square

3 Construction of simple bipartite graphs

Till now, `ConstructBipartiteRealization()` generates any bipartite realization of the bipartite degree sequence A . But, it is easy to modify `AddVertex()` so that the output of `ConstructBipartiteRealization()` is always a simple graph. One obvious condition can be stated as follows.

(a) If the Algorithm is inserting the j^{th} edge of vertex v_s (with $j > 1$ and $v_s \in V_2$) and v_k ($v_k \in V_1$) is already adjacent to v_s , then no more available edge incident to v_k should be chosen. This would prevent `ConstructBipartiteRealization()` from outputting graphs with multiple edges (v_s, v_k) . Thus this condition is necessary, but it is not sufficient. Indeed, it is easy to see that the following must also apply.

(b) While inserting vertex v_s and avoiding choosing edges incident to v_k so as not to construct multiple edges (v_s, v_k) , `ConstructBipartiteRealization()` may fall into a stage where there are more edges incident to v_k than there are vertices left to insert, and G , the graph output by `ConstructBipartiteRealization()` would then have a multiple edge (v_1, v_k) .

(c) Let A_1 and A_2 be (separately) ordered in non decreasing order, where a_1 is the largest entry of A_1 and a_{n_1+1} is the largest entry of A_2 . Let M_k be the set of the last k entries of A_1 and let $\max(k) = a_{n_1-k+1}$. Let there be an entry a_s in A_2 satisfying the following.

(f1). $s - n_1 \geq \max(k)$, (i.e., the number of entries of A_2 preceding a_s is

greater or equal to the maximal entry in M_k)

(f2). $\alpha_s > n_1 - k$, (i.e., inserting v_s would require more neighbours than there are vertices in $V_1 \setminus M_k$) and,

f(3).

$$\sum_{j=n_1+1}^{s-1} \alpha_j \geq \sum_{i=k}^{n_1} \alpha_i + \sum_{i=1}^{k-1} \max(0, \alpha_i - n + s).$$

(that is, the number of edges required to insert vertices of V_2 prior to v_s exceeds the number of edges available on vertices in M_k plus the minimum number of edges that a vertex v_i (with $v_i \in V_1 \setminus M_k$) has to concede prior to the s^{th} iteration to prevent v_i from having more edges than there are vertices left to be inserted from the s^{th} iteration onwards.)

If $\alpha_s \in A_2$ satisfies (f1), (f2) and (f3), then α_s is said to be k -fat. Let F_k denote the set of all the entries that are k -fat. See an illustration in Figure 4.

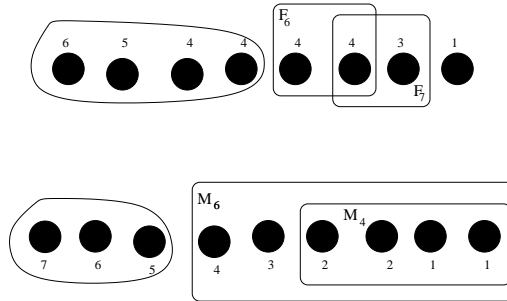


Figure 4: $A = (7, 6, 5, 4, 3, 2, 2, 1, 1 : 6, 5, 4, 4, 4, 3, 1)$, where $A_1 = (7, 6, 5, 4, 3, 2, 2, 1, 1)$ and $A_2 = (6, 5, 4, 4, 4, 4, 3, 1)$. Entries are labelled so that the leftmost entry of A_1 is α_1 and the rightmost entry of A_2 is α_{17} . The entries α_{14} and α_{15} are 6-fat while α_{15} and α_{16} are 7-fat.

Now, if (a) is to be respected and `ConstructBipartiteRealization()` chose every vertex in M_k to concede an edge to every one of the $s - n_1$ vertices preceding v_s , then `ConstructBipartiteRealization()` would get stuck at the stage of inserting vertex v_s . This is because by (f1) and (f3), no vertex in M_k would have any edge to concede to v_s and so there would be a maximum of $n_1 - k$ available vertices. But by (f2), vertex v_s needs more adjacent neighbors than the only $n_1 - k$ available vertices. Hence, `ConstructBipartiteRealization()` must take some precautionary measures by not exhausting all the edges incident to vertices in M_k prior to the insertion of v_s .

Figure 5 illustrates how the Algorithm would get stuck at its s^{th} iteration.

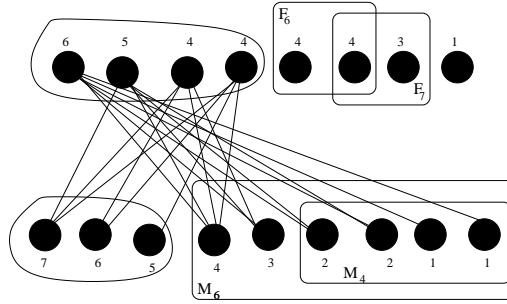


Figure 5: This is a choice of edges that may exhaust all the edges incident to vertices in M_6 prior to the 14th iteration. In this choice, vertex v_1, v_2 and v_3 must concede 3, 2 and 1 edges respectively lest they would have too many edges after the 13th iteration. Still, vertex v_{14} would not get inserted fully and the Algorithm would stall.

Although (a), (b) and (c) seem to contradict each other, this section defines all these conditions in a formal settings and proves that they can be satisfied simultaneously. Although the analysis seems lengthy, this set of conditions are just inequalities involving the number of edges and vertices already inserted and the number of edges and vertices left to be inserted at each stage of the Algorithm. Moreover, checking these conditions at each iteration of `AddVertex()` requires checking $\mathcal{O}(n^2)$ inequalities altogether. Thus it does not add to the running time.

Let $A = (A_1 : A_2)$ be a bipartite degree sequence of a simple graph, where A_1 and A_2 have respectively n_1 and n_2 entries such that $n_1 + n_2 = n$. We recall that E_{av} represents the set of available edges. That is, edges that are incident to v_1 and vertices inserted before the n_1^{th} iteration of `ConstructBipartiteRealization()`, that is, the vertices of V_1 . For $v_j \in V_1$, we recall that E_{v_j} is the set of available edges incident to v_j . That is, the set of parallel edges connecting v_1 and v_j . Obviously $E_{v_j} \subseteq E_{av}$ for all j . In particular, E_{v_1} is the set of loops incident to v_1 .

Some of the Algorithms given in the literature, such as in [1], have the disadvantage that it has to restart. The algorithm given here allows to choose only edges such that it never has to restart. In order to be able to do that, the choice of edges at every stage must be such that no vertex is incident to too many edges of the 'wrong type'.

If at its s^{th} iteration, Algorithm `ConstructBipartiteRealization()` is inserting the vertex v_s that has degree a_s , then `ConstructBipartiteRealization()` has to

call the routine `AddVertex()` that has to go through α_s iterations. We recall that the $(s, t)^{\text{th}}$ stage of `ConstructBipartiteRealization()` is the iteration where `AddVertex()` inserts the t^{th} edge of the s^{th} vertex. Let $X_{s,t}$ and $|X|_{s,t}$ denote respectively a set and its cardinality at the $(s, t)^{\text{th}}$ stage of `ConstructBipartiteRealization()`.

To help the reader, we first introduce the motivation for the definitions. At each stage of constructing a simple graph, every vertex v_j , where $v_j \in V_1$, must be connected by at most one edge to any other v_k , where $v_k \in V_2$. So, if some vertex v_j has more available edges than the vertices left to be inserted after its s^{th} iteration, `ConstructBipartiteRealization()` would never be able to get rid of all these multiple edges, which would then appear in the final graph. This prompts the following definitions. The vertex v_j where $j \leq n_1$ (i.e., $v_j \in V_1$) is *due* if

$$|E_{v_j}|_{st} = n - (s - 1), \quad (1)$$

that is, E_{v_j} has as many edges as there are vertices left to be inserted. The vertex v_j is *overdue* if

$$|E_{v_j}|_{st} > n - (s - 1), \quad (2)$$

that is, there are too many available edges incident to v_j and whatever are the future choices, the Algorithm would never output a simple graph. The vertex v_j is *undue* if it is neither due nor overdue. Obviously, a stage is *due*, *undue*, *overdue* if there is a vertex that is due, undue or overdue, respectively.

Let M_k be the set of the last k entries of A_1 . An entry α_s in A_2 is *k-fat* if conditions (f1), (f2) and (f3) are satisfied. We let F_k to denote the set of vertices that are *k-fat*. A bipartite degree sequence A is *fat* if it contains a *k-fat* entry for some integer $k > 0$.

Let $r_i = \alpha_i - n_1 + k$, where k is the largest integer such that α_i is *k-fat*. The $(s, t)^{\text{th}}$ stage is *ruined* if there is an entry α_i with $i > s$ (that is, the vertex v_i is not inserted yet) that is fat and the number of vertices in M_k that are available is less than r_i . It is *not ruined* otherwise.

The next lemma indicates that once `ConstructBipartiteRealization()` has taken a ‘wrong path’, it is impossible to mend the situation.

Lemma 1 *Suppose `ConstructBipartiteRealization()` is inserting the vertex v_s such that $s > n_1$, (i.e., inserting v_s into V_2). Then the following hold.*

(a) *If the vertex v_j is due, it is due or overdue at the next stage. If it is overdue, it is overdue at any future stage.*

(b) If the $(s, t)^{\text{th}}$ stage is overdue, then the previous stage (the stage inserting the previous edge) is either due or overdue.

(c) If the $(s, t)^{\text{th}}$ stage is ruined, then the next stage is also ruined.

Proof.

(a) Suppose v_j is due and $\text{Addvertex}()$ does not choose an edge from E_{v_j} . Since no edge of E_{v_j} is chosen, the left side of Equation 1 remains same while the right hand side either goes down by one if $\text{ConstructBipartiteRealization}()$ moves to a new vertex v_{s+1} or stays the same if $\text{ConstructBipartiteRealization}()$ moves to another edge $t+1$ of the same vertex v_s . Hence the next stage is due or overdue. On the other hand, if $\text{Addvertex}()$ chooses an edge from E_{v_j} , the left hand side goes down by 1 and the right one stays the same. But if E_{v_j} concedes only one edge to v_s (as we shall see shortly), E_{v_j} is still due at the insertion of vertex v_{s+1} . Similar arithmetical arguments as above show that if v_j is overdue, it stays overdue.

(b) Suppose v_j is overdue at the $(s, t)^{\text{th}}$ stage but is undue at the stage inserting the previous edge. Then at the previous stage, we have

$$|E_{v_j}| < n - (s - 1). \quad (3)$$

Now, either the last edge inserted is chosen from E_{v_j} or not. Moreover, in either case, Algorithm $\text{ConstructBipartiteRealization}()$ moves to a new vertex or not. If it stays on the same vertex and the chosen edge is not from E_{v_j} , the right and the left hand sides of Equation 3 are both unchanged. Hence v_j is undue at the $(s, t)^{\text{th}}$ stage, which is a contradiction. If it stays on the same vertex and the chosen edge is from E_{v_j} , the left hand side of Equation 3 goes down by 1 while the right hand side is unchanged. Hence v_j is also undue at the $(s, t)^{\text{th}}$ stage and this is again a contradiction.

Suppose $\text{ConstructBipartiteRealization}()$ moves to a new vertex. If the chosen edge is not from E_{v_j} , the right hand side of Equation 3 goes down by 1 while the right hand side is unchanged. Hence v_j is due at the $(s, t)^{\text{th}}$ stage, a contradiction. If the chosen edge is from E_{v_j} , both left hand and right hand sides of Equation 3 go down by 1. Hence v_j is normal at the $(s, t)^{\text{th}}$ stage, a contradiction.

(c) Assume that the $(s, t)^{\text{th}}$ stage is ruined. That is, there is a fat vertex v_i that is not inserted yet, but the number of vertices in M_k which are available is less than r_i . But, at the next stage, this number can never increase. Thus it would also be ruined. \square

While Lemma 1 says that once `ConstructBipartiteRealization()` takes a wrong path, it is impossible to mend it, the next routine gives preventive measures to avoid getting into that wrong path in the first place.

ChooseCorrectEdge()

Let A be not fat and `ConstructBipartiteRealization()` is at its $(s, t)^{\text{th}}$ stage with $s > n_1$ (that is, inserting vertex v_s into V_2). Then,

(1) for each vertex $v_j \in V_1$, do not choose an edge in E_{v_j} if there is already an edge (v_s, v_j) .

(2) if the vertex v_j is due, pick an edge from E_{v_j} . If many vertices are due, pick an edge uniformly at random from the vertices that are due.

Now assume that A is fat and for some integer $k > 0$, F_k is not empty. Then, for every entry $a_i \in F_k$ choose at random $r_i = a_i - n_1 + k$ different entries in M_k . The only condition imposed on the choice is that an entry a_j can be chosen at most once for each fat vertex and at most a_j times for all the fat vertices combined. If a_i is k -fat, let R_i , called the *reserve pool* of a_i , be the set of vertices in M_k chosen for a_i . Let R_{ij} , the *reserve matrix*, be an n_1 by n_2 matrix whose columns are indexed from 1 to n_1 (indices of entries of A_1), and rows are indexed from $n_1 + 1$ to n (indices of entries of A_2), and $R_{ij} = 1$ if the entry $a_j \in R_i$, and zero otherwise. Obviously, the sum of entries in row i is equal to r_i and the sum of entries of column j must be less or equal to a_j . At the $(s, t)^{\text{th}}$ stage, a vertex $v_j \in V_1$ is *exhausted* if the sum of row j plus the number of vertices adjacent to v_j equals a_j . (that is, the number of edges already conceded by v_j and the number of edges of v_j in the reserve pools equals a_j).

(3) If `ConstructBipartiteRealization()` is at its $(s, t)^{\text{th}}$ stage with $s > n_1$ and a_s is not fat, then apply (1) and (2) subject to not choosing a vertex v_j if v_j is exhausted. If a_s is fat, first choose all the vertices in R_s , then apply (1) and (2) if necessary.

Complexity Issues

Before proving that the conditions set in routine `ChooseCorrectEdge()` are necessary and sufficient to sample a simple bipartite graph at random, we observe that, if $A = (a_1, a_2, \dots, a_n) = (A_1 : A_2)$ where A_1 and A_2 have respectively n_1 and n_2 entries such that $n_1 + n_2 = n$ and $\sum_{i=1}^n a_i = 2m$, `ChooseCorrectEdge()` runs altogether in $\mathcal{O}(n_1 n_2)$ steps. Indeed, at the s^{th} iteration of `ConstructBipartiteRealization()`, `ChooseCorrectEdge()` has to check Equation 1 only once

for every vertex $v_j \in V_1$. But there are n_2 iterations and n_1 vertices v_j with $j \leq n_1$. This takes $\mathcal{O}(n_1 n_2)$ steps. Constructing the Reserve Matrix R requires $\mathcal{O}(n_1 n_2)$ steps as one has to check Conditions (f1), (f2) and (f3) for each of the n_2 entries of A_2 and writing the $n_1 n_2$ entries of the matrix R .

Theorem 3 *Algorithm ConstructBipartiteRealization() reconstructs a simple graph if and only if AddVertex() calls the routine ChooseCorrectEdge(). In other words, ConstructBipartiteRealization() outputs a simple graph if and only if the choice of edges satisfies Conditions (1), (2) and (3).*

Proof. Assume to the contrary that Conditions (1) and (2) hold but ConstructBipartiteRealization() outputs a bipartite graph G with multiple edges or loops. By Condition (1) there can not be a multiple edge connecting two vertices v_j and v_k such that $j \leq n_1$ and $k > n_1$. Moreover, by the definition of the routine ConstructBipartiteRealization(), there can not be a double edge (v_k, v_l) where $k, l > n_1$. Hence if G fails to be a simple graph, it must have either a loop or a multiple edge incident to v_1 and v_j such that $j \leq n_1$.

So, in G , let the vertex v_1 is incident to either a loop e or a multiple edge (v_1, v_j) such that $j \leq n_1$. But, by the definition of the bipartition, the number of edges incident to v_1 at the end of the n_1^{th} iteration of ConstructBipartiteRealization() equals the number of edges left to be inserted until ConstructBipartiteRealization() terminates. Hence, some vertex v_k such that $k > n_1$ is not fully inserted. This is a contradiction.

Conversely, let the condition (1) or (2) be not satisfied and let G be the realization output by ConstructBipartiteRealization(). If condition (1) is not satisfied at the $(s, t)^{\text{th}}$ stage, this would create a double edge (v_j, v_s) with $j \leq n_1$ and $s > n_1$. Now, since Algorithm Addvertex() can not concede the double edge (v_j, v_s) anymore as they are lost, the double edge (v_j, v_s) would appear in G . Hence G would not be simple. Assume that the condition (2) is not satisfied. That is, there is a vertex v_j with $j \leq n_1$ that is due at the $(s, t)^{\text{th}}$ stage, where $s > n_1$, but Algorithm Addvertex() does not pick any of the elements of E_{v_j} for all the remaining edges conceded to v_s . Then v_j is overdue at the insertion of vertex v_{s+1} , and by Lemma 1(b) it remains overdue until the end of Algorithm 2. Hence G is not simple as it must have a multiple edge (v_i, v_j) . If condition (3) is not satisfied, Algorithm ConstructBipartiteRealization() may stall. \square

Let a *correct edge* and *vertex* be an edge chosen by Algorithm ChooseCorrectEdge and a vertex incident to a correct edge, respectively. So if ConstructBipartiteRealization() terminates, we have shown that it always outputs a

simple graph. It remains to show that it always terminates by showing that there is always a correct edge so that conditions (1) and (2) can be satisfied at every stage of `ConstructBipartiteRealization()`.

Theorem 4 *Algorithm `ConstructBipartiteRealization()` always terminates. That is, Conditions (1) and (2) are always satisfied at every stage of `ConstructBipartiteRealization()`.*

Proof. Suppose A does not contain any fat entry. That is, as long as an edge $e = (v_1, v_j)$ is a correct vertex, it can be chosen. Obviously, Condition (1) can always be forced on `AddVertex()`. But, while trying hard to satisfy Condition (1), the algorithm may let a vertex v_j of V_1 , to become overdue. If at the $(s, t)^{\text{th}}$ stage the vertex v_j is due, we prove that it is always possible to concede an edge from E_{v_j} to v_s .

So assume to the contrary that v_j is due but `Addvertex()` can not pick an edge from E_{v_j} . This is possible only if there are too many vertices that are due. That is, $\alpha_s < n'_1 \leq n_1$, where n'_1 is the number of vertices that are due at the $(s, t)^{\text{th}}$ stage. But we also have $\alpha_s \geq \alpha_{s+1} \geq \dots \geq \alpha_n$. Moreover, as all these n'_1 vertices are due, each of them is incident to $n - s$ available edges. Hence we have $\alpha_s + \alpha_{s+1} + \dots + \alpha_n < n'_1(n - s)$. That is, there are more available edges than there are edges left to be inserted until `ConstructBipartiteRealization()` terminates. This contradicts Observation 2.

Let the entry α_i be k -fat. If all the correct edges $e = (v_1, v_j)$ such that $v_j \in M_k$ are conceded prior to the insertion of the vertex v_i , then by definition of fat entry, Algorithm `ConstructBipartiteRealization()` would stall as there would not be enough edges to connect to v_i . But, we assume that the Algorithm reserved r_i edges to concede to v_i . Hence v_i can always be inserted. So, we only need to check (c1), whether putting some edges in reserve would prevent some non-fat vertex v_s from being inserted for lack of correct edges and, (c2), whether it is always possible to construct the reserve matrix R_{ij} .

(c1) Assume that $s < i$. That is, v_s precedes v_i . Let all vertices preceding v_s have been inserted but there are not enough correct edges to insert v_s . This is possible if reserving edges for vertices in F_k and inserting vertices preceding v_s exhausts q vertices of V_1 and $\alpha_s > n_1 - q$. Without loss of generality, we may assume that the last q vertices of V_1 are exhausted. So, let the available vertices be vertices v_1, \dots, v_{n_1-q+1} . If the number of available edges is less than α_s , then $A_1 < A_2$. This is a contradiction. So, let the number of available edges be greater or equal to α_s . Thus the number of available vertices is less than α_s , so that Condition (1) prevents α_s edges from being connected to v_s . Let H

be the graph obtained after the insertion of v_{s-1} by 'fully' connecting all the vertices in $V_2 \setminus v_s$, making sure to connect vertices in F_k with edges that are reserved for them in R_{ij} . Then, by the definition of r_i , it is easy to check that every vertex in F_k is adjacent to every vertex in $V_1 \setminus M_k$. Also, since all the vertices in M_q are exhausted after the insertion of v_{s-1} , one can check that none of the vertices in $M_q \setminus M_k$ is adjacent to a vertex in $V_2 \setminus (F_k \cup V_{\leq s})$, where $V_{\leq s}$ denotes the set of vertices from v_{n_1+1} up to v_s . (i.e., $V_2 \setminus (F_k \cup V_{\leq s})$ is the set of vertices between v_s and F_k). Thus, only the vertices in $V_1 \setminus M_q$ are adjacent to vertices in $V_2 \setminus (F_k \cup V_{\leq s})$. Since all the vertices, except for v_s are properly connected and $|A_1| = |A_2|$, the number of available edges is a_s but the number of available vertices is less than a_s . Therefore, by the pigeonhole principle, there is an available vertex having at least two available edges. Without loss of generality, we may consider v_1 to be the culprit.

Now, since only the vertices in $V_1 \setminus M_q$ are adjacent to the vertices in $V_2 \setminus (F_k \cup V_{\leq s})$, either v_1 is adjacent to all the vertices in $V_2 \setminus (F_k \cup V_{\leq s})$ or it is not. If it is, then v_1 was due during an iteration prior to or during the insertion of v_{s-1} and the algorithm did not select it to concede an edge. This is a contradiction. Suppose that it is not adjacent to some vertex $v_t \in V_2 \setminus (F_k \cup V_{\leq s})$. Then $a_t < n_1 - q$, since v_t is fully connected. But, by the non decreasing ordering of A_2 , we also have $a_t \geq a_i$. Moreover, since $a_i \in F_k$, we have $a_i \geq n_1 - k$. Hence we have $a_i \geq n_1 - k > n_1 - q > a_t$. This is also a contradiction. Therefore vertex v_s can be fully inserted. See Figure 6 which helps to understand notations in part (c1).

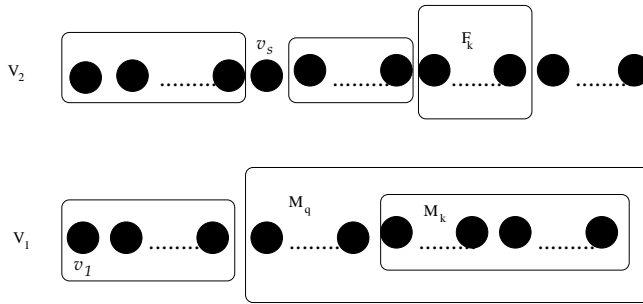


Figure 6:

Finally, let a_i be k -fat, a_s be not k -fat and $s > i$. (that is, v_s is to be inserted after v_i). If there are not enough correct edges to connect to v_s , then $|A_1| < |A_2|$. This is a contradiction.

(c2) Suppose that it is not possible to build the reserve matrix. But, since

$a_i \leq n_1$ for all entries in F_k , this would imply either $\sum_{F_k} a_i > \sum_{V_1 \setminus M_k} a_i + \sum_{M_k} a_i = |A_1|$, or $a_i > n_1$ for some entry $a_i \in F_k$. This is a contradiction. \square

It still remains to show that the algorithm constructs all the simple realizations of A .

Lemma 2 *Let G_{n_1, n_2} be the n_1, n_2 -complete bipartite graph. That is, the bipartite graph where one part contains n_1 vertices each having degree n_2 and the second part contains n_2 vertices each of degree n_1 . Then ConstructBipartiteRealization() satisfying Conditions (1) and (2) can reconstruct G_{n_1, n_2} as a realization of $A = (A_1 : A_2)$ where A_1 has n_1 entries $a_i = n_2$ and A_2 has n_2 entries $a_j = n_1$.*

Proof. At the beginning of the $(n_1 + 1)^{\text{th}}$ iteration, $E_{v_j} = n_1$ for each of the n_1 vertices already inserted. Hence each such vertex is due. Now, the vertex v_{n_1+1} has degree $a_{n_1+1} = n_1$ by the definition of A . Hence, by Condition (2), AddVertex() chooses one edge from each of the n_1 vertices v_j with $j \leq n_1$ and inserts v_{n_1+1} completely. By Lemma 1, each v_j is still due at the $(n_1 + 2)^{\text{th}}$ iteration. Again, by Condition (2), AddVertex() chooses one edge from each of the n_1 vertices v_j with $j \leq n_1$ and inserts v_{n_1+2} completely. And so on, until the insertion of vertex v_n , and Algorithm ConstructBipartiteRealization() outputs the graph G_{n_1, n_2} . \square

Let G be a graph, a *delete-minor* of $G' = G \setminus e$ is the graph obtained from G by deleting the edge e . If $A = (A_1 : A_2)$ is a bipartite degree sequence, let A' be the degree sequence obtained from A by subtracting 1 from two of its entries a_i and a_j , where $a_i \in A_1$ and $a_j \in A_2$. Thus, if A is the degree sequence of a bipartite graph G , then A' is the degree sequence of some delete-minor of G .

Lemma 3 *If ConstructBipartiteRealization() satisfying Conditions (1) and (2) can reconstruct G as a realization of A , then it can reconstruct all the delete-minors of G that are realizations of A' .*

Proof. Let G be a bipartite graph output by Algorithm ConstructBipartiteRealization() and let $G \setminus e$ be a delete-minor of G . In the graph G , let the edge e be incident to vertices v_j and v_k having respectively degrees a_j and a_k , where $j \leq n_1$ and $k > n_1$. Thus in $G \setminus e$, vertices v_j and v_k have degrees $a_j - 1$ and $a_k - 1$. Let f be any edge of $G \setminus e$. Since G is output by ConstructBipartiteRealization(), there is a series of choices of correct edges such that f can be

inserted. In that series of choices either e is inserted before or after f . If e is inserted after f , the same series of choices would insert f in $G \setminus e$. If e is inserted before f , the same series of choices, minus the insertion of e , would also lead to the insertion of f in $G \setminus e$, since Algorithm `ConstructBipartiteRealization()` does not need to insert any edge incident to v_j and v_k as their degrees are down by 1. \square

Corollary 1 *Let G be a simple bipartite realization of a degree sequence $A = (A_1 : A_2)$ where A_1 and A_2 have n_1 and n_2 entries respectively. Then there is a positive probability that G is output by Algorithm `ConstructBipartiteRealization()` if Conditions (1) and (2) are satisfied.*

Proof. Every simple bipartite graph having one part of n_1 vertices and another of n_2 vertices can be obtained from G_{n_1, n_2} by a series of deletions. \square

3.1 Sampling all bipartite realizations uniformly

Although Theorem 2 shows that the routine `ConstructBipartiteRealization()` can construct a realization of A in time linear on the number of edges of its realizations, we need the next result to show that it can construct any bipartite realization of A with equal probability, provided we define the probability $\pi(e)$ with which `AddVertex()` has to insert the edge e . If at its k^{th} iteration `ConstructBipartiteRealization()` is to insert the vertex v_k that has degree a_k , then `ConstructBipartiteRealization()` has to call `AddVertex()` that has to go through a_k iterations. Let the $(s, t)^{\text{th}}$ stage of `ConstructBipartiteRealization()` be the iteration where `AddVertex()` inserts the t^{th} edge of the s^{th} vertex and let $G^{(s, t)}$ denote the graph obtained at that $(s, t)^{\text{th}}$ stage. With this notation, let $G^{(s)}$ be the graph $G^{(s, a_s)}$. The *random reconstruction tree*, denoted by \mathcal{T} , is a directed rooted tree where the root is the sole realization of the degree sequence $A^{(1)}$, and the $(s, t)^{\text{th}}$ level contains all those possible graphs obtainable after inserting the t^{th} edge of the s^{th} vertex, and there is an arc from a graph H at level i to the graph G at level $i + 1$ if it is possible to move from H to G by the concession of a single available edge. Realizations of A are thus the leaves of the tree \mathcal{T} . With this formalism, sampling a random bipartite realization of the degree sequence A is equivalent to performing a random walk from the root until a leaf is reached, and every step of the random walk consists of walking along a random arc of \mathcal{T} . See Figure 7 for an illustration.

Rejection sampling

Let G be a realization of A . That is, G is a leaf of the tree \mathcal{T} . Obviously, there are many paths of \mathcal{T} leading to G . Let p be such a path and let $\pi_p(G)$ denote the probability to reach G along the path p . Now $\pi_p(G)$ can easily be computed on the fly since $\pi_p(G) = \prod_{e \in E(G)} \pi(e)$, where $E(G)$ denotes the set of edges of G and $\pi(e)$ is the probability to choose the edge e . Now $\pi(e) = \frac{1}{|V_{\text{cor}}|}$, where V_{cor} is the set of all correct vertices at the insertion of e . The only problem is that G can be reached from many paths. The next result proves that all these paths have equal probability.

Lemma 4 *Let G be a realization of A that can be reached through the paths p and q of \mathcal{T} . Then $\pi_p(G) = \pi_q(G)$.*

Proof. Let $E(G)$ denote the set of edges of G . Then, p can be seen as a re-ordering of a subset of edges chosen along q . Now, since the vertices are added in the same order along q as along p , we may only consider the case where p and q differ on a single vertex and edges e and f are interchanged in p and q . Let $V_{\text{cor}}(e)$ and $V_{\text{cor}}(f)$ denote the sets of correct vertices at the insertion of e and f , respectively. If the Algorithm can choose either the edge e or f , then $V_{\text{cor}}(e) = V_{\text{cor}}(f)$ and the probability to choose either must be the same. \square

Lemma 4 allows to compute $\pi(G)$ on the fly. For any path p leading to G , we have

$$\pi(G) = \prod_{e \in G} \pi(e) = \prod_{e \in G} \frac{1}{|V_{\text{corr}}(e)|},$$

where $V_{\text{corr}}(e)$ is the set of vertices in V_1 that are incident to some correct edge. Hence, to get $\pi(G)$ on the fly, one set $\pi(G) = \pi(G^{n_1}) = 1$. For every partial realization $G^{(i)}$ from (G^{n_1}) to G multiply $\pi(G)$ by $\frac{1}{|V_{\text{corr}}(e)|}$. Finally output $\pi(G)$ with G . Now let $\min(\pi)$ be a lower bound of the probabilities to reach of the realizations of A . This lower bound can be calculated using only parameters of A . Indeed, if $|V_{\text{av}}(e)|$ stands for the number of vertices in V_1 that are adjacent to v_1 at the insertion of edge e , then we have the inequality $\frac{1}{|V_{\text{av}}(e)|} \leq \frac{1}{|V_{\text{corr}}(e)|} \leq \pi(e)$ and, for any realization G , we have

$$\prod_{e \in G} \frac{1}{|V_{\text{av}}(e)|} \leq \prod_{e \in G} \pi(e) \leq \pi(G).$$

Finally, since $|V_{v_1}(e)| \leq n_1$ and every realization of A has m edges, we get

$$\frac{1}{n_1^m} \leq \prod_{e \in G} \frac{1}{|V_{\text{av}}(e)|} \leq \prod_{e \in G} \pi(e) \leq \pi(G).$$

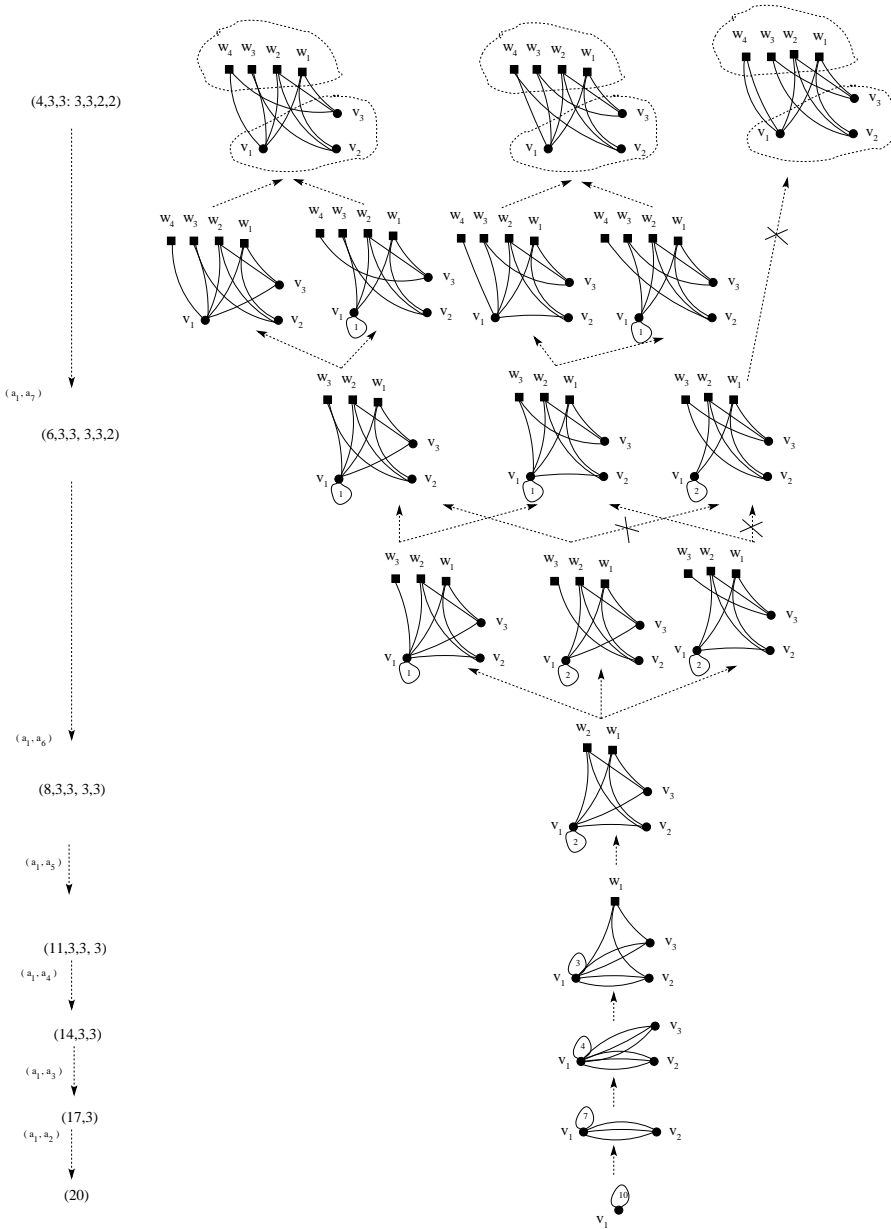


Figure 7: Random reconstruction tree of $(4, 3, 3 : 3, 3, 2, 2)$. The level of \mathcal{T} on the same height as the degree sequence $A^{(i)}$ corresponds to all the graphs having $A^{(i)}$ as their degree sequence. The arrows that are crossed denote the edges that would not lead to a simple realization.

Algorithm RejectionSampling()

Input: Bipartite degree sequence $A = (A_1 : A_2)$, where A_1 and A_2 have n_1 and n_2 entries respectively such that $n_1 + n_2 = n$ and an integers r_1 .

Output: A sequence of r_1 bipartite simple realizations of A where every realization has equal probability.

Step 1 Put A_1 and A_2 in non decreasing order.

Step 2 Construct the recursion chain of A by calling the routine ConstructBipartiteRecursionChain().

Step 3 Call ConstructBipartiteRealization() to construct the realization G . Let $\pi(G)$ be the probability computed on the fly and get u , a random number in $(0, 1)$. If $u < \frac{\min(\pi)}{\pi(G)}$, accept G and go back to Step 3 until one gets r_1 realizations. Else, reject G and go back to Step 3 until one gets r_1 realizations.

Obviously, Algorithm RejectionSampling() samples every realization of A with the same probability equal to $\min(\pi)$. Now, it is known that Step 1 takes $\log(n_1) + \log(n_2)$ iterations and, as shown earlier, Step 2 takes $n_1 + n_2$ iterations. In Step 3, ChooseCorrectEdge() does $\mathcal{O}(n_1 n_2)$ inequality checks altogether while AddVertex() needs $2m$ iterations to insert all the vertices. Thus, the overall running time to get the minimum probability is given by

$$\log(n_1) + \log(n_2) + r_1(n_1 n_2 + 2m) = \mathcal{O}(r_1(n_1 n_2 + 2m)) \asymp \mathcal{O}(3r_1 m) \asymp \mathcal{O}(m).$$

Finally, T , the running time of generating a realization of A uniformly, is a geometric random variable with expected running time given by $\frac{1}{\pi(\text{acc})}$ where $\pi(\text{acc})$ is the acceptance probability for the realization G with the highest probability of being output by ConstructBipartiteRealization(). So

$$\pi(\text{acc}) = \frac{\min(\pi)}{\pi(G)} = \frac{\min(\pi)}{\prod_{e \in G} |V_{\text{corr}}(e)|}.$$

Now if $n_2 \rightarrow \infty$, then $|V_{\text{corr}}(e)| \rightarrow \frac{n_1}{2}$ on average. Therefore,

$$\pi(\text{acc}) \rightarrow \frac{\min(\pi)}{(\frac{2}{n_1})^m} = \frac{\frac{1}{n_1^m}}{(\frac{2}{n_1})^m} = \frac{1}{2^m}.$$

Hence $T \rightarrow 2^m$. For the typical Darwin tables m is about 40 edges. Thus 2^m is a manageable running time.

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