Finding Complete Bipartite Subgraphs in Bipartite Graphs

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Finding complete bipartite subgraphs in
Bipartite Graphs*

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Abstract

Given a bipartite graph G, we investigate the problem of determining whether it contains a K_{a,b} - a complete bipartite graph with a union of two
node sets and a common vertex set. In the former setting, we solve the problem in time O(a^3 b), for a = O(b^2) - a number of edges of G and b in
O(|V|). In the latter setting, we solve the problem in time O(a^3 b^3), for a = O(b^3) where |V| is the number of edges of G and b is the number of
nodes in one of the node sets.

1 Introduction

Given a graph G, a natural question is what is the size of the largest complete bipartite subgraph H of the graph? It is well known that a complete bipartite subgraph of a graph G is a cycle if G contains a cycle of certain length, i.e., a < b, or a clique of certain size, i.e., a ≥ b.

In this paper we present an algorithm that tests efficiently whether a bipartite graph G contains a cycle of length a small than b or contains a clique of size a smaller than b. For the former setting, the algorithm runs in time \(O(n^{3/2} \sqrt{a})\), where n is the number of edges of G. For the latter setting, the algorithm runs in time \(O(n^{3/2} \sqrt{a})\), where n is the number of edges of G.

The method we use is similar to the one used in [1]. We transform the graph and look at it as a geometrical setting, where it becomes a pattern recognition problem. The set of points in the plane determines whether the set contains a certain number of points that form the vertices of a quadrilateral. For more on our setting, see [1].

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Figure 1: Transforming a bipartite graph to a set of points. Observe that the graph contains a $K_{2,3}$ (the thick edges) and hence the set of points contains a $2 \times 3$ subgrid (the thick points).

The problem lies in the fact that a set of points cannot contain too many subgrids of size $l \times l$ without containing a subgrid of size $l \times m$.

The sequel of this paper is organised as follows.

First we give the transformation from the graph problem to the geometrical setting and we give an algorithm that counts the number of $l \times l$ subgrids in a set of points. Then we show how to use this algorithm to determine whether there is a $l \times m$ subgrid.

2 The algorithm

Let $G = (V \cup W, E)$ be a bipartite graph. The nodes in $V$ are labeled $v_1, \ldots, v_{|V|}$ and the nodes in $W$ are labeled $w_1, \ldots, w_{|W|}$. The edge between $v_i$ and $w_j$ is denoted $(v_i, w_j)$. With this graph $G$ we associate a set $S_G$ of points in the plane as follows: $S_G = \{(i, j) | (v_i, w_j) \in E\}$. Thus there is a one-to-one correspondence between the edges of $G$ and the points in $S_G$, and two points in $S_G$ have the same $x$-coordinate (y-coordinate) if and only if the corresponding edges are incident to the same node of $V$ ($W$). Hence, under this transformation a $K_{t,m}$ corresponds to a set of points that lie on an axis-parallel grid of size $l \times m$ (see Figure 1). A $K_{2,2}$, for example, corresponds to four points that are the vertices of an axis-parallel rectangle. This transformation is also used in [4].
In the remainder of this paper we will take the geometrical point of view and show how to determine whether a set $S$ of $n$ points in the plane contains a subset of size $l \times m$ whose points lie on an axis-parallel grid of size $l \times m$. We denote such a grid simply by $K_{l,m}$.

2.1 Counting $K_{l,i}$'s

We first turn our attention to the counting (and reporting) of the $K_{l,i}$'s in a set $S$ of points in the plane. The idea is to partition $S$ into subsets whose points have equal first coordinates and distinguish between small subsets and large subsets. Let a column (of size $l$) be a set of $l$ points that lie on a common vertical line; similarly, a row is a set of points that lie on a common horizontal line. We identify a column $C$ consisting of the points $(x,y_1),\ldots,(x,y_l)$, where $y_1 < y_2 < \cdots < y_l$, with the point $(y_1,\ldots,y_l)$ in $l$-dimensional space. Thus to find a $K_{l,i}$ we have to look for $l$ columns that are identical (under this identification). Because of the size of the small subsets we can afford to enumerate all columns of size $l$ in the small subsets.

Then we search with these columns in the large subsets. To do this efficiently we make a distinction between large subsets that are relatively small and large subsets that are relatively large. This way we count the number of times such a column is present. Then we can calculate the number of $K_{l,i}$'s. The only $K_{l,i}$'s that we have not counted so far are those having all their columns in large subsets. These $K_{l,i}$'s are counted by reversing the roles of $x$- and $y$-coordinates.

We next give a more detailed description of this algorithm.

1. Partition $S$ into subsets whose points have equal first coordinate. Let $S_1,\ldots,S_a$ be the subsets of size at most $n^\alpha$ (the small subsets). Let $L_1,\ldots,L_b$ be the sets of size between $n^\alpha$ and $n^\beta$, and let $L'_1,\ldots,L'_{b'}$ be the sets of size greater than $n^\beta$ (the large subsets that are relatively small and the ones that are relatively large). $\alpha$ and $\beta$ are parameters to be chosen later (with $\beta \geq \alpha$).

2. For each $S_i$, enumerate all possible columns of size $l$ having the points in $S_i$. Let $C_1,\ldots,C_d$ be the different columns thus obtained. (Recall that we identify columns with points in $l$-dimensional space and that two columns are equal if the set of $y$-coordinates of the points in one column equals the set of $y$-coordinates of the points in the other column.) Store the multiplicity of the columns (the number of times a column occurs) in an array $M_S[1..d]$. E.g., if some $C_i$ occurs in $S_j$ and $S_j'$ but in no other small subset we set $M_S[i] := 2$. The multiplicities are found by sorting the columns lexicographically. Now columns that are identical are adjacent in this sorted list.

3. Search for the occurrences in the large subsets of columns found in step 2 in the small subsets as follows. Initialize all entries in an array $M_L[1..d]$ to 0.
(i) Count the number of times a $C_i$ occurs in an $L_j$ as follows. Enumerate all columns in every $L_j$ and sort them lexicographically. Now walk simultaneously along this sorted list and the sorted list of $C_i$'s. If a $C_i$ occurs in $c$ $L_j$’s, set $M_L[i] := c$.

(ii) Count the number of times a $C_i$ occurs in an $L_j'$ as follows. Build a search tree $T$ on the $y$-coordinates of the points in $\bigcup_{1 \leq j \leq v} L_j'$. Store at each leaf $\gamma$ in $T$ a list containing all $j$–values such that $L_j'$ contains a point whose $y$–coordinate is equal to $\gamma$, the $y$–coordinate corresponding to $\gamma$. With this tree $T$ we can find in $O(\log n)$ time a list of all $L_j'$’s that contain a point with some specific $y$–coordinate by searching in $T$ with this $y$–coordinate.

For each $C_i$ do the following. For every coordinate $y$ of $C_i$ (these are the $y$–coordinates of the points in the column) search in $T$ and mark each $L_j'$ that contains a point whose $y$–coordinate is equal to $y$. (The names of these subsets are stored at the leaf $\gamma$ with $\gamma_y = y$.) Check each $L_j'$ to see if it is marked $l$ times and if this is the case (every point of $C_i$ has been found in $L_j'$, i.e., $C_i$ occurs in $L_j'$) increment $M_L[i]$.

4. To count the number of $K_{l,d}$’s with all columns in large subsets we partition $\bigcup_{1 \leq j \leq l} L_j \cup \bigcup_{1 \leq j \leq v} L_j'$, the set of all points in large subsets, into new subsets whose points have equal $y$–coordinate. We then enumerate for each new subset all possible rows having their points in this subset and we compute $M'[1 \ldots d']$, an array containing the multiplicities of the $d'$ different rows, by sorting the rows (as $M$ was computed in step 2).

5. The number of $K_{l,d}$’s is now given by

$$\sum_{i=1}^{d} \left( M_S[i] \cdot \sum_{j=1}^{d'} \left( \frac{M[i]}{M_L[i]} \right) - \sum_{i=1}^{d} \left( \frac{M[i]}{M_L[i]} \right) \right),$$

where we define $\binom{r}{t} = 0$ for $r < t$. (The last summation gives the number of $K_{l,d}$’s having all columns in large subsets which have a corresponding column in a small subset. These $K_{l,d}$’s are counted twice, so we have to subtract this number.)

This leads to:

**Lemma 1** The number of $K_{l,d}$’s in a set of $n$ points can be counted in time $O(n^{\frac{d}{t+1} + \frac{d}{n-1}})$.

**Proof:** We first note that all the coordinates of the points lie in the range $1 \ldots n$. Hence, the sorting steps that are needed in the algorithm only take time linear in the number objects to be sorted (plus $O(n)$, of course). Thus the first step of the
The algorithm only takes $O(n)$ time and the second step takes time linear in the number of columns in small subsets, which is

$$
\sum_{i=1}^{a} \left( \frac{|S_i|}{l} \right) \leq \sum_{i=1}^{a} |S_i|^l \leq (n^\alpha)^{l-1} \sum_{i=1}^{a} |S_i| \leq n^{al-\alpha+1}.
$$

Step 2(i) takes time linear in the number of columns in large subsets that are relatively small plus the number of columns in the small subsets. Hence, this takes $O(n^{\beta l-\beta+1})$ time.

Step 2(ii) takes time $O(\sum_{i=1}^{d} l(\log n + n^{1-\beta}))$ (with every point in each $C_i$ we have to search, which takes $O(\log n)$ time, and increment at most $n^{1-\beta}$ counters) plus $O(n)$ (to build the tree). Because $d = O(n^{al-\alpha+1})$, the time for the second step is $O(n^{2+\alpha(l-1)-\beta})$ (plus $O(n)$).

The time spent in the third step is bounded by $O(n^{\alpha(1-l)+l})$, the maximal number of rows in large subsets.

Hence, to obtain the best time bound we have to minimize

$$\max(n^{al-\alpha+1}, n^{\beta l-\beta+1}, n^{2+\alpha(l-1)-\beta}, n^{\alpha(1-l)+l}).$$

Note that the first term is never greater than the second term since $\beta \geq \alpha$. Also observe that if we choose $\alpha = \frac{1}{2} - \epsilon$ and $\beta = \frac{1}{2} + \epsilon$ the second and fourth term are equal. If we let $\epsilon = \frac{1}{4l-2}$, then we even have that the second and fourth term are equal to the third. Since the second term increasing in $\beta$, the third term is decreasing in $\beta$ and increasing in $\alpha$ and the fourth term is decreasing in $\alpha$, this must be the optimal solution. Substituting the values $\alpha = \frac{1}{2} - \frac{1}{4l-2}$ and $\beta = \frac{1}{2} + \frac{1}{4l-2}$ yields the claimed time bound. \qed

2.2 Finding $K_{l,m}$’s

We will now show how to use the algorithm of the previous section to determine efficiently whether a set of points contains a $K_{l,m}$. Assume w.l.o.g. that $l \leq m$.

**Lemma 2** If a set $S$ of $n$ points in the plane contains more than $2\binom{m-1}{l} n^{l+\frac{1}{2}}$ $K_{l,m}$’s then $S$ contains at least one $K_{l,m}$.

**Proof:** Suppose that $S$ contains no $K_{l,m}$. Partition $S$ into subsets whose points have equal $x$-coordinate. In other words, two points are in the same subset iff they have the same $x$-coordinate. Distinguish between subsets that contain at most $\sqrt{n}$ points (the small subsets) and subsets that contain more than $\sqrt{n}$ points (the large subsets).

Let $S_1, \ldots, S_a$ be the collection of small subsets. Consider all columns of size $l$ in the small subsets. Clearly, the number of these columns is

$$
\sum_{i=1}^{a} \left( \frac{|S_i|}{l} \right) \leq \sum_{i=1}^{a} |S_i|^l \leq \sqrt{n^{l-1}} \sum_{i=1}^{a} |S_i| \leq n^{\frac{l}{2}+\frac{1}{2}}.
$$

5
Now the number of occurrences of some particular column is smaller than \( m \) (otherwise there would be a \( K_{l,m} \)). Thus any particular column can take part in at most \( \binom{m-1}{l} \) \( K_{l,t} \)'s. Since there were at most \( n^{\frac{1}{2}+\frac{t}{2}} \) columns in the small subsets, there are no more than \( \binom{m-1}{l} n^{\frac{1}{2}+\frac{t}{2}} \) \( K_{l,t} \)'s having at least one column in a small subset.

To count the number of \( K_{l,t} \)'s having all their columns in large subsets, we note that there cannot be more than \( \sqrt{n} \) large subsets. Hence, the number of rows in large subsets is bounded by \( n^{\frac{1}{2}+\frac{t}{2}} \). The same argument now shows that there are at most \( \binom{m-1}{l} n^{\frac{1}{2}+\frac{t}{2}} \) \( K_{l,t} \)'s having all rows, and thus all columns, in large subsets. \( \Box \)

Our strategy to decide upon the existence of a \( K_{l,m} \) will thus be to count the number of \( K_{l,t} \)'s (where \( l \leq m \)). If this number is greater than \( 2 \binom{m-1}{l} n^{\frac{1}{2}+\frac{t}{2}} \) then we know by the above lemma that there must also be a \( K_{l,m} \). If the number is smaller then we sort all the \( K_{l,t} \)'s we have found lexicographically; a \( K_{l,m} \) would give \( m-l+1 \) \( K_{l,t} \)'s that are adjacent in this sorted list. (Observe that the algorithm that counts the \( K_{l,t} \)'s can easily be adapted to report the \( K_{l,t} \)'s. Thus what we do is start reporting them until their number becomes too large. Then we know that there must be a \( K_{l,m} \). If this does not happen we have all \( K_{l,t} \)'s available for the last step.)

The counting of the \( K_{l,t} \)'s is done using the algorithm presented in the previous section. We now describe the last step (determining whether there is a \( K_{l,m} \) if the number of \( K_{l,t} \)'s is small) in a little more detail. Let \( K \) be a \( K_{l,t} \). Let \( x_1, \ldots, x_l \) and \( y_1, \ldots, y_l \) be the different \( x \)-coordinates resp. \( y \)-coordinates of the points of \( K \) in sorted order. We identify \( K \) with the point \((x_1, \ldots, x_l, y_1, \ldots, y_l)\) in \( 2l \)-dimensional space. This induces an ordering on the \( K_{l,t} \)'s that corresponds to the lexicographical ordering on the corresponding points in \( 2l \)-dimensional space. This is easily seen that if there is a 'vertical' \( K_{l,m} \) (a \( K_{l,m} \) with \( l \) columns and \( m \) rows) then there must be at least \( m-l+1 \) \( K_{l,t} \)'s that are adjacent in the above ordering whose points together form a \( K_{l,m} \). (Notice that not every \( K_{l,m} \) has this property. However, if there are vertical \( K_{l,m} \)'s, then at least one has.) Thus by walking along the ordered list of \( K_{l,t} \)'s we can check for the existence of a 'vertical' \( K_{l,m} \) in time linear in the length of this list. To find 'horizontal' \( K_{l,m} \)'s we just reverse the roles of the \( x \)- and \( y \)-coordinates.

This leads to:

**Theorem 1** Given a bipartite graph \( G \) with \( n \) edges, it can be decided in time \( O(n^{\frac{1}{2}+\frac{t}{2}-\frac{m-1}{2}}) \) whether \( G \) contains a \( K_{l,m} \) as a subgraph, where \( t = \min(l, m) \).

**Proof:** The fact that the algorithm presented above correctly solves the (equivalent geometrical) problem immediately follows from Lemma 2. Let \( l \leq m \). The time taken by the algorithm that counts the number of \( K_{l,t} \)'s is \( O(n^{\frac{1}{2}+\frac{t}{2}-\frac{m-1}{2}}) \) (Lemma 1) and the time needed to test if there is a \( K_{l,m} \) when the number of \( K_{l,t} \)'s is smaller
than $2 \left( \frac{m-1}{t} \right) n^{\frac{1}{t}+\frac{1}{2}}$ is linear in this number since, as we already noted, the sorting takes only linear time. □

References


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Abstract

Given a bipartite graph $G$, we investigate the problem of determining whether $G$ contains a $K_{l,m}$ (a complete bipartite graph with $l$ nodes in one node set and $m$ nodes in the other node set) as a subgraph. Using a transformation to a geometrical setting, we solve this problem in time $O(n^{l+\frac{1}{2}-\frac{1}{l(t+1)}})$, where $n$ is the number of edges of $G$ and $t = \min(l, m)$.

1 Introduction

Given a graph $G$, a natural question to ask is whether $G$ contains some specific graph $H$ as a subgraph. One could, for example, ask whether $G$ contains a cycle of certain length ([1, 3, 6, 7]) or a clique of certain size ([1, 5]).

In this paper we present an algorithm that tests efficiently whether a bipartite graph $G$ contains a $K_{l,m}$ as a subgraph. (A $K_{l,m}$ is a complete bipartite graph with $l$ nodes in one node set and $m$ nodes in the other node set.) For variable $l$ and $m$ this problem is NP-complete (see Garey and Johnson [2], problem GT24), so we consider $l$ and $m$ to be constants. For the case $l = m = 2$ efficient algorithms already exist; Chiba and Nishizeki [1] and van Kreveld and de Berg [4] give algorithms that work in time $O(n\sqrt{n})$, where $n$ is the number of edges in $G$. (In fact, Chiba and Nishizeki solve the more general problem of finding cycles of length four in an arbitrary graph.)

Our new algorithm works for arbitrary $l$ and $m$ and takes time $O(n^{\frac{1}{l(t+1)}-\frac{1}{l(t+1)}})$, where $t = \min(l, m)$. For $l = m = 2$ this is only slightly worse than the previous results.

The method we use is similar to the one used in [4]. We transform the graph problem to a geometrical setting, where it becomes a pattern recognition problem: Given a set of points in the plane, determine whether this set contains a subset of points that form the vertices of an $l \times m$ subgrid. The key to our solution to this

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problem lies in the fact that a set of points cannot contain too many subgrids of size $l \times l$ without containing a subgrid of size $l \times m$.

The sequel of this paper is organised as follows.

First we give the transformation from the graph problem to the geometrical setting and we give an algorithm that counts the number of $l \times l$ subgrids in a set of points. Then we show how to use this algorithm to determine whether there is a $l \times m$ subgrid.

2 The algorithm

Let $G = (V \cup W, E)$ be a bipartite graph. The nodes in $V$ are labeled $v_1, \ldots, v_{|V|}$ and the nodes in $W$ are labeled $w_1, \ldots, w_{|W|}$. The edge between $v_i$ and $w_j$ is denoted $(v_i, w_j)$. With this graph $G$ we associate a set $S_G$ of points in the plane as follows: $S_G = \{(i,j)|(v_i, w_j) \in E\}$. Thus there is a one-to-one correspondence between the edges of $G$ and the points in $S_G$, and two points in $S_G$ have the same $x$-coordinate ($y$-coordinate) if and only if the corresponding edges are incident to the same node of $V$ ($W$). Hence, under this transformation a $K_{l,m}$ corresponds to a set of points that lie on an axis-parallel grid of size $l \times m$ (see Figure 1). A $K_{2,2}$, for example, corresponds to four points that are the vertices of an axis-parallel rectangle. This transformation is also used in [4].
In the remainder of this paper we will take the geometrical point of view and show how to determine whether a set $S$ of $n$ points in the plane contains a subset of size $l \times m$ whose points lie on an axis-parallel grid of size $l \times m$. We denote such a grid simply by $K_{l,m}$.

2.1 Counting $K_{l,l}$'s

We first turn our attention to the counting (and reporting) of the $K_{l,l}$'s in a set $S$ of points in the plane. The idea is to partition $S$ into subsets whose points have equal first coordinates and distinguish between small subsets and large subsets. Let a column (of size $l$) be a set of $l$ points that lie on a common vertical line; similarly, a row is a set of points that lie on a common horizontal line. We identify a column $C$ consisting of the points $(x,y_1),\ldots,(x,y_l)$, where $y_1 < y_2 < \cdots < y_l$, with the point $(y_1,\ldots,y_l)$ in $l$-dimensional space. Thus to find a $K_{l,l}$ we have to look for $l$ columns that are identical (under this identification). Because of the size of the small subsets we can afford to enumerate all columns of size $l$ in the small subsets. Then we search with these columns in the large subsets. To do this efficiently we make a distinction between large subsets that are relatively small and large subsets that are relatively large. This way we count the number of times such a column is present. Then we can calculate the number of $K_{l,l}$'s. The only $K_{l,l}$'s that we have not counted so far are those having all their columns in large subsets. These $K_{l,l}$'s are counted by reversing the roles of $x$- and $y$-coordinates.

We next give a more detailed description of this algorithm.

1. Partition $S$ into subsets whose points have equal first coordinate. Let $S_1,\ldots,S_n$ be the subsets of size at most $n^\alpha$ (the small subsets). Let $L_1,\ldots,L_k$ be the sets of size between $n^\alpha$ and $n^\beta$, and let $L'_1,\ldots,L'_{\nu}$ be the sets of size greater than $n^\beta$ (the large subsets that are relatively small and the ones that are relatively large). $\alpha$ and $\beta$ are parameters to be chosen later (with $\beta \geq \alpha$).

2. For each $S_i$, enumerate all possible columns of size $l$ having the points in $S_i$. Let $C_1,\ldots,C_d$ be the different columns thus obtained. (Recall that we identify columns with points in $l$-dimensional space and that two columns are equal if the set of $y$-coordinates of the points in one column equals the set of $y$-coordinates of the points in the other column.) Store the multiplicity of the columns (the number of times a column occurs) in an array $M_S[1..d]$. E.g., if some $C_i$ occurs in $S_j$ and $S_j'$ but in no other small subset we set $M_S[i] := 2$. The multiplicities are found by sorting the columns lexicographically. Now columns that are identical are adjacent in this sorted list.

3. Search for the occurrences in the large subsets of columns found in step 2 in the small subsets as follows. Initialize all entries in an array $M_L[1..d]$ to 0.
(i) Count the number of times a \( C_i \) occurs in an \( L_j \) as follows. Enumerate all columns in every \( L_j \) and sort them lexicographically. Now walk simultaneously along this sorted list and the sorted list of \( C_i \)'s. If a \( C_i \) occurs in \( c L_j \)'s, set \( M_{L[i]} := c \).

(ii) Count the number of times a \( C_i \) occurs in an \( L_j' \) as follows. Build a search tree \( T \) on the \( y \)-coordinates of the points in \( \bigcup_{1 \leq j \leq \nu} L_j' \). Store at each leaf \( \gamma \) in \( T \) a list containing all \( j \)-values such that \( L_j' \) contains a point whose \( y \)-coordinate is equal to \( \gamma \), the \( y \)-coordinate corresponding to \( \gamma \). With this tree \( T \) we can find in \( O(\log n) \) time a list of all \( L_j' \)'s that contain a point with some specific \( y \)-coordinate by searching in \( T \) with this \( y \)-coordinate.

For each \( C_i \) do the following. For every coordinate \( y \) of \( C_i \) (these are the \( y \)-coordinates of the points in the column) search in \( T \) and mark each \( L_j' \) that contains a point whose \( y \)-coordinate is equal to \( y \). (The names of these subsets are stored at the leaf \( \gamma \) with \( \gamma_y = y \).) Check each \( L_j' \) to see if it is marked \( l \) times and if this is the case (every point of \( C_i \) has been found in \( L_j' \), i.e., \( C_i \) occurs in \( L_j' \)) increment \( M_{L[i]} \).

4. To count the number of \( K_{l,t} \)'s with all columns in large subsets we partition \( \bigcup_{1 \leq j \leq \delta} L_j \cup \bigcup_{1 \leq j \leq \nu} L_j' \), the set of all points in large subsets, into new subsets whose points have equal \( y \)-coordinate. We then enumerate for each new subset all possible rows having their points in this subset and we compute \( M'[1...d'] \), an array containing the multiplicities of the \( d' \) different rows, by sorting the rows (as \( M \) was computed in step 2).

5. The number of \( K_{l,t} \)'s is now given by

\[
\sum_{i=1}^{d} \left( M_{S[i]} \right) + M_{L[i]} + \sum_{j=1}^{d'} \left( M'[j] \right) - \sum_{i=1}^{d} \left( M_{L[i]} \right),
\]

where we define \( \binom{r}{t} = 0 \) for \( r < t \). (The last summation gives the number of \( K_{l,t} \)'s having all columns in large subsets which have a corresponding column in a small subset. These \( K_{l,t} \)'s are counted twice, so we have to subtract this number.)

This leads to:

**Lemma 1** The number of \( K_{l,t} \)'s in a set of \( n \) points can be counted in time \( O(n^{\frac{4}{l+1}} \frac{4}{l} - n^{\frac{4}{l+1}}) \).

**Proof:** We first note that all the coordinates of the points lie in the range 1 \( \ldots \) \( n \). Hence, the sorting steps that are needed in the algorithm only take time linear in the number objects to be sorted (plus \( O(n) \), of course). Thus the first step of the
algorithm only takes \( O(n) \) time and the second step takes time linear in the number of columns in small subsets, which is

\[
\sum_{i=1}^{a} \binom{|S_i|}{l} \leq \sum_{i=1}^{a} |S_i|^l \leq (n^\alpha)^l \sum_{i=1}^{a} |S_i| \leq n^{\alpha l^{-\alpha+1}}.
\]

Step 2(i) takes time linear in the number of columns in large subsets that are relatively small plus the number of columns in the small subsets. Hence, this takes \( O(n^{\beta l^{-\beta+1}}) \) time.

Step 2(ii) takes time \( O(n^{d \log n + n^{1-\beta}}) \) (with every point in each \( C_i \) we have to search, which takes \( O(n \log n) \) time, and increment at most \( n^{1-\beta} \) counters) plus \( O(n) \) (to build the tree). Because \( d = O(n^{\alpha l^{-\alpha+1}}) \), the time for the second step is \( O(n^{2\alpha (l^{-1})^{-\beta}}) \) (plus \( O(n) \)).

The time spent in the third step is bounded by \( O(n^{\alpha (1-l)^{l+l}}) \), the maximal number of rows in large subsets.

Hence, to obtain the best time bound we have to minimize

\[
\max(n^{\alpha l^{-\alpha+1}}, n^{\beta l^{-\beta+1}}, n^{2\alpha (l^{-1})^{-\beta}}, n^{\alpha (1-l)^{l+l}}).
\]

Note that the first term is never greater than the second term since \( \beta \geq \alpha \). Also observe that if we choose \( \alpha = \frac{1}{2} - \varepsilon \) and \( \beta = \frac{1}{2} + \varepsilon \) the second and fourth term are equal. If we let \( \varepsilon = \frac{1}{\sqrt{l-2}} \), then we even have that the second and fourth term are equal to the third. Since the second term increasing in \( \beta \), the third term is decreasing in \( \beta \) and increasing in \( \alpha \) and the fourth term is decreasing in \( \alpha \), this must be the optimal solution. Substituting the values \( \alpha = \frac{1}{2} - \frac{1}{\sqrt{l-2}} \) and \( \beta = \frac{1}{2} + \frac{1}{\sqrt{l-2}} \) yields the claimed time bound.

2.2 Finding \( K_{l,m} \)'s

We will now show how to use the algorithm of the previous section to determine efficiently whether a set of points contains a \( K_{l,m} \). Assume w.l.o.g. that \( l \leq m \).

Lemma 2 If a set \( S \) of \( n \) points in the plane contains more than \( 2C_{l-1}m \) \( K_{l,m} \)'s then \( S \) contains at least one \( K_{l,m} \).

Proof: Suppose that \( S \) contains no \( K_{l,m} \). Partition \( S \) into subsets whose points have equal \( x \)-coordinate. In other words, two points are in the same subset iff they have the same \( x \)-coordinate. Distinguish between subsets that contain at most \( \sqrt{n} \) points (the \( small \) subsets) and subsets that contain more than \( \sqrt{n} \) points (the \( large \) subsets).

Let \( S_1, \ldots, S_a \) be the collection of small subsets. Consider all columns of size \( l \) in the small subsets. Clearly, the number of these columns is

\[
\sum_{i=1}^{a} \binom{|S_i|}{l} \leq \sum_{i=1}^{a} |S_i|^l \leq \sqrt{n}^{l-1} \sum_{i=1}^{a} |S_i| \leq n^{l^{-l+1}}.
\]
Now the number of occurrences of some particular column is smaller than \( m \) (otherwise there would be a \( K_{l,t} \)). Thus any particular column can take part in at most \( \binom{m-1}{l} \) \( K_{l,t} \)'s. Since there were at most \( n^{\frac{3}{4}+\frac{1}{4}} \) columns in the small subsets, there are no more than \( \binom{m-1}{l} n^{\frac{3}{4}+\frac{1}{4}} \) \( K_{l,t} \)’s having at least one column in a small subset.

To count the number of \( K_{l,t} \)'s having all their columns in large subsets, we note that there cannot be more than \( \sqrt{n} \) large subsets. Hence, the number of rows in large subsets is bounded by \( n^{\frac{3}{4}+\frac{1}{4}} \). The same argument now shows that there are at most \( \binom{m-1}{l} n^{\frac{3}{4}+\frac{1}{4}} \) \( K_{l,t} \)’s having all rows, and thus all columns, in large subsets.

\( \Box \)

Our strategy to decide upon the existence of a \( K_{l,m} \) will thus be to count the number of \( K_{l,t} \)'s (where \( l \leq m \)). If this number is greater than \( 2 \left( \binom{m-1}{l} n^{\frac{3}{4}+\frac{1}{4}} \right) \), then we know by the above lemma that there must also be a \( K_{l,m} \). If the number is smaller then we sort all the \( K_{l,t} \)'s we have found lexicographically; a \( K_{l,m} \) would give \( m - l + 1 \) \( K_{l,t} \)'s that are adjacent in this sorted list. (Observe that the algorithm that counts the \( K_{l,t} \)'s can easily be adapted to report the \( K_{l,t} \)'s. Thus what we do is start reporting them until their number becomes too large. Then we know that there must be a \( K_{l,m} \). If this does not happen we have all \( K_{l,t} \)'s available for the last step.)

The counting of the \( K_{l,t} \)'s is done using the algorithm presented in the previous section. We now describe the last step (determining whether there is a \( K_{l,m} \) if the number of \( K_{l,t} \)'s is small) in a little more detail. Let \( K \) be a \( K_{l,t} \). Let \( x_1, \ldots, x_l \) and \( y_1, \ldots, y_{l} \) be the different \( x \)-coordinates resp. \( y \)-coordinates of the points of \( K \) in sorted order. We identify \( K \) with the point \((x_1, \ldots, x_l, y_1, \ldots, y_l)\) in \( 2l \)-dimensional space. This induces an ordering on the \( K_{l,t} \)'s that corresponds to the lexicographical ordering on the corresponding points in \( 2l \)-dimensional space. It is easily seen that if there is a ‘vertical’ \( K_{l,m} \) (a \( K_{l,m} \) with \( l \) columns and \( m \) rows) then there must be at least \( m - l + 1 \) \( K_{l,t} \)'s that are adjacent in the above ordering whose points together form a \( K_{l,m} \). (Notice that not every \( K_{l,m} \) has this property. However, if there are vertical \( K_{l,m} \)'s, then at least one has.) Thus by walking along the ordered list of \( K_{l,t} \)'s we can check for the existence of a ‘vertical’ \( K_{l,m} \) in time linear in the length of this list. To find ‘horizontal’ \( K_{l,m} \)'s we just reverse the roles of the \( x \)- and \( y \)-coordinates.

This leads to:

**Theorem 1** Given a bipartite graph \( G \) with \( n \) edges, it can be decided in time \( O(n^{\frac{3}{4}+\frac{1}{4}} - n^{\frac{1}{4}}) \) whether \( G \) contains a \( K_{l,m} \) as a subgraph, where \( t = \min(l, m) \).

**Proof:** The fact that the algorithm presented above correctly solves the (equivalent geometrical) problem immediately follows from Lemma 2. Let \( l \leq m \). The time taken by the algorithm that counts the number of \( K_{l,t} \)'s is \( O(n^{\frac{3}{4}+\frac{1}{4}} - n^{\frac{1}{4}}) \) (Lemma 1) and the time needed to test if there is a \( K_{l,m} \) when the number of \( K_{l,t} \)'s is smaller
than $2\left( m^{-1} \right) n^{\frac{1}{2} + \frac{1}{2}}$ is linear in this number since, as we already noted, the sorting takes only linear time.

References


