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Maximizing Several Cuts Simultaneously

DANIELA KÜHN and DERYK OSTHUS

School of Mathematics, Birmingham University, Edgbaston, Birmingham B15 2TT, UK
(e-mail: {kuehn,osthus}@maths.bham.ac.uk)

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Consider two graphs G_1 and G_2 on the same vertex set V and suppose that G_i has m_i edges. Then there is a bipartition of V into two classes A and B so that, for both $i = 1, 2$, we have $e_{G_i}(A, B) \geqslant m_i/2 - \sqrt{m_i}$. This gives an approximate answer to a question of Bollobás and Scott. We also prove results about partitions into more than two vertex classes. Our proofs yield polynomial algorithms.

1. Introduction

Given a graph G with m edges, the Max-Cut problem is to determine (the size of) the maximum cut in G. For complete graphs, the largest cut has size $m/2 + o(m)$. On the other hand, it is well known that a cut of size at least $m/2$ in a graph G can be found using the natural greedy algorithm. Improving this, Edwards [6, 7] showed that every graph with m edges has a cut of size

$$m/2 + \sqrt{\frac{m}{8} + \frac{1}{64}} - \frac{1}{8},$$

which is best possible. The Max-Cut problem is equivalent to finding a bipartition V_1, V_2 of the vertex set of G which minimizes $e_G(V_1) + e_G(V_2)$, where $e_G(V_i)$ denotes the number of edges in the subgraph of G induced by V_i. The problem when one is looking for a partition into k classes $V_1, \ldots, V_k$ which minimizes all $e_G(V_i)$ simultaneously, *i.e.*, which minimizes $\max\{e_G(V_1), \ldots, e_G(V_k)\}$, was studied by Bollobás and Scott [2, 3, 5] as well as Porter [12, 13, 14]; see also [4] for a survey. Partition problems concerning graphs of large minimum degree were investigated in [9, 16], for example.

Here, we suppose that we are given several graphs on the same vertex set and we want to find a bipartition which maximizes the sizes of the cuts for all these graphs simultaneously. This problem was posed by Bollobás and Scott [5]. More precisely, they asked the following question: What is the largest integer $f(m)$ such that, whenever G_1 and G_2 are two graphs with m edges on the same vertex set V, there exists a bipartition

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of V in which, for both $i = 1, 2$, at least $f(m)$ edges of G_i go across (*i.e.*, their endvertices lie in different partition classes)? They suggested that perhaps even $f(m) = (1 - o(1))m/2$, *i.e.*, that we can do almost as well as in the case where we only have a single graph. Theorem 1.1 shows that this is indeed the case.

Given a graph G and disjoint subsets A, B of its vertex set, let $e_G(A, B)$ denote the number of edges between A and B.

Theorem 1.1. *Consider graphs $G_1, \dots, G_\ell$ on the same vertex set V and suppose that G_i has m_i edges. Then there is a bipartition of V into two classes A and B so that for all $i = 1, \dots, \ell$ we have*

$$e_{G_i}(A, B) \geqslant \frac{m_i}{2} - \sqrt{\ell m_i/2}.$$

Rautenbach and Szigeti [15] observed that, even for $\ell = 2$, we cannot ensure that $e_{G_i}(A, B) \geqslant m_i/2$ for all i. Indeed, let G_1 and G_2 be two edge-disjoint cycles of length 5 on the same vertex set. (So $G_1 \cup G_2 = K_5$.) They also proved that $f(m) \geqslant m/2 - \Delta^3$ if $\Delta(G_i) \leqslant \Delta$ for $i = 1, 2$. (This answers the problem of Bollobás and Scott if $(\Delta(G_i))^3 = o(m)$ for $i = 1, 2$.)

The following result for partitions of graphs into more than two parts shows that, simultaneously for all graphs, we can ensure that the number of crossing edges is almost as large as one would expect in a random partition (and almost the value one can ensure if one were to partition only a single graph).

Theorem 1.2. *Let $k \geqslant 2$. Consider graphs $G_1, \dots, G_\ell$ on the same vertex set V and suppose that G_i has m_i edges. Then there is a partition of V into k classes $V_1, \dots, V_k$ so that, for all $i = 1, \dots, \ell$, the number of edges spanned by the k-partite subgraph of G_i induced by $V_1, \dots, V_k$ is at least*

$$\frac{(k-1)m_i}{k} - \sqrt{2\ell m_i/k}.$$

In fact, if $\Delta(G_i) = o(m_i)$ for each i, then we can strengthen the conclusion. The next theorem shows that there is a partition of V into k classes where each of the $\binom{k}{2}$ bipartite graphs spanned by two of the partition classes contains almost $2m_i/k^2$ edges for all $i = 1, \dots, \ell$ simultaneously. Again, this is about the number of edges one would expect in a random partition.

Theorem 1.3. *Let $k \geqslant 2$ and $0 < \varepsilon \leqslant 1/(9\ell^2 k^4)$. Consider graphs $G_1, \dots, G_\ell$ on the same vertex set V. Suppose that G_i has m_i edges and that $\Delta(G_i) \leqslant \varepsilon m_i$ for all $i = 1, \dots, \ell$. Then there is a partition of V into k classes $V_1, \dots, V_k$ so that, for all $i = 1, \dots, \ell$ and for all s, t with $1 \leqslant s < t \leqslant k$, we have*

$$e_{G_i}(V_s, V_t) \geqslant \frac{2m_i}{k^2} - \varepsilon^{1/4} m_i$$

and

$$e_{G_i}(V_s) \geqslant \frac{m_i}{k^2} - \varepsilon^{1/4} m_i.$$

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Note that, even for $\ell = 1$, the condition that $\Delta(G_i) \leqslant \varepsilon m_i$ cannot be omitted completely. For example, the result is obviously false if G is a star. On the other hand, a result of Bollobás and Scott [5, Theorem 3.2] implies that in the case when the maximum degree of each G_i is bounded by a constant Δ, the bound on $e_{G_i}(V_s, V_t)$ in Theorem 1.3 can be improved to $2m_i/k^2 - C$, where $C = C(\ell, \Delta)$ (and similarly for $e_{G_i}(V_s)$). Note that this implies that if G has bounded maximum degree, then one can achieve a bounded error term in Theorems 1.1 and 1.2 as well.

The proofs of Theorems 1.1–1.3 can be derandomized to yield polynomial time algorithms which find the desired partitions (see Section 4).

2. An open problem

Consider an r-uniform hypergraph $\mathscr{H}$ with m hyperedges. It is easy to see that there is a partition $V_1, \ldots, V_r$ of the vertex set of $\mathscr{H}$ such that at least $r!m/r^r$ hyperedges of $\mathscr{H}$ meet every V_i (in other words, each r-uniform hypergraph contains an r-partite subhypergraph with at least $r!m/r^r$ hyperedges). To verify this, consider the expected number of hyperedges which meet every V_i in a random partition of the vertices.

As observed by Keevash and Sudakov (personal communication), Theorem 1.1 does not carry over to r-uniform hypergraphs $\mathscr{H}_1, \ldots, \mathscr{H}_\ell$: one cannot always find a partition of the vertex set into r classes such that in each hypergraph at least $r!m_i/r^r - o(m_i)$ hyperedges meet all the r partition classes (where m_i denotes the number of hyperedges of $\mathscr{H}_i$). Indeed, consider for example the case when $r = 3$ and $\ell = 6$. Let V be a set of $n \geqslant 4$ vertices and fix four vertices $v_1, \ldots, v_4$ in V. For all pairs $1 \leqslant i < j \leqslant 4$ consider the hypergraph $\mathscr{H}_{ij}$ on V whose hyperedges are precisely those triples in V which contain both v_i and v_j. For any partition of V into three classes, one class contains at least two of the vertices $v_1, \ldots, v_4$. If v_i and v_j are such vertices, then no hyperedge of $\mathscr{H}_{ij}$ meets all the three partition classes.

However, it is still an open problem whether there is also a counterexample which uses only two hypergraphs.

Problem 2.1. *Suppose that $\mathscr{H}_1$ and $\mathscr{H}_2$ are r-uniform hypergraphs on the same vertex set V such that $\mathscr{H}_i$ has m_i hyperedges. Does there always exist a partition of V into r classes $V_1, \ldots, V_r$ such that, for both $i = 1, 2$, at least $r!m_i/r^r - o(m_i)$ hyperedges of $\mathscr{H}_i$ meet each of the classes $V_1, \ldots, V_r$?*

Given an r-uniform hypergraph $\mathscr{H}$ and distinct vertices $x, y \in \mathscr{H}$, denote by $N_{\mathscr{H}}(x, y)$ the number of hyperedges containing both x and y. Let $\Delta_2(\mathscr{H})$ denote the maximum of $|N_{\mathscr{H}}(x, y)|$ over all pairs $x \neq y$. One can adapt our proof of Theorem 1.3 to show that Problem 2.1 has an affirmative answer in the case when $\Delta_2(\mathscr{H}_i) = o(m_i)$ for each i (even for several hypergraphs instead of two). We omit the details.

3. Proofs

The proofs all proceed by considering a random partition and analysing this using the second moment method.

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Lemma 3.1. *Let $c \in \mathbb{R}$ with $c > 1/2$. Suppose that G is a graph with m edges whose vertex set is V. Consider a random bipartition of V into two classes A and B which is obtained by including each $v \in V$ into A with probability $1/2$ independently of all other vertices in V. Then, with probability at least $1 - 1/(2c)$, we have*

$$e_G(A,B) \geqslant \frac{m}{2} - \sqrt{cm/2}.$$

If we apply the above result with $c = \ell$ (say) to the graphs in Theorem 1.1, the failure probability for each of them is less than $1/(2\ell)$. Summing up all these failure probabilities immediately implies Theorem 1.1.

Proof of Lemma 3.1. For every edge e of the graph G, define an indicator variable X_e as follows: if one endvertex of e is in A and the other is in B, then let $X_e := 1$; otherwise let $X_e := 0$. Clearly, $\mathbb{P}[X_e = 1] = 1/2$. Also, for $e, e' \in E(G)$ with $e \neq e'$, we have

$$\mathbb{E}[X_e \cdot X_{e'}] = \mathbb{P}[X_e = 1, X_{e'} = 1] = \frac{1}{2}\mathbb{P}[X_e = 1 \mid X_{e'} = 1] = \frac{1}{4}.$$

Note that the final equality holds regardless of whether or not e and e' have an endvertex in common. Now let $X := \sum_{e \in E(G)} X_e$. Thus X counts the number of edges between A and B and $\mathbb{E}X = m/2$. Let

$$\sum_{\substack{e,e' \in E(G) \\ e \neq e'}}$$

denote the sum over all ordered pairs e, e' of distinct edges in G. Then, using the fact that $\mathbb{E}[X_e^2] = \mathbb{E}[X_e]$, we have

$$\begin{aligned}\mathbb{E}[X^2] &= \sum_{e \in E(G)} \mathbb{E}[X_e] + \sum_{\substack{e,e' \in E(G) \\ e \neq e'}} \mathbb{E}[X_e \cdot X_{e'}] \\ &= \mathbb{E}[X] + \sum_{\substack{e,e' \in E(G) \\ e \neq e'}} \frac{1}{4} = \frac{m}{2} + \frac{m(m-1)}{4} = \frac{m(m+1)}{4}.\end{aligned}$$

This in turn implies that the variance of X satisfies $\mathrm{Var}X = \mathbb{E}[X^2] - (\mathbb{E}X)^2 = m/4$. The result now follows from a straightforward application of Chebyshev's inequality:

$$\mathbb{P}[X \leqslant m/2 - \sqrt{cm/2}] \leqslant \mathbb{P}[|X - \mathbb{E}X| \geqslant \sqrt{cm/2}] \leqslant \frac{2\mathrm{Var}X}{cm} = \frac{1}{2c}. \qquad \square$$

Proof of Theorem 1.2. As in Lemma 3.1, we first consider a single graph G with m edges and vertex set V. Consider a random partition of V into k disjoint sets V_j, which is obtained by including each $v \in V$ into V_j with probability $1/k$ independently of all other vertices. Let $X_e := 0$ if the edge e has both its endpoints in some V_j and let $X_e := 1$ otherwise. So $\mathbb{P}[X_e = 1] = (k-1)/k$. Also, it is easy to check that $\mathbb{E}[X_e \cdot X_{e'}] = (k-1)^2/k^2$. Again, this holds regardless of whether or not e and e' have an endvertex in common. Let X denote the number of edges whose endvertices lie in different vertex classes. Thus

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$\mathbb{E}X = \frac{k-1}{k}m$ and

$$\mathbb{E}[X^2] = \sum_{e\in E(G)} \mathbb{E}[X_e] + \sum_{\substack{e,e'\in E(G)\\ e\neq e'}} \mathbb{E}[X_e \cdot X_{e'}]$$
$$= \frac{k-1}{k}m + m(m-1)\frac{(k-1)^2}{k^2} \leqslant \frac{m}{k} + (\mathbb{E}[X])^2.$$

Therefore $\mathrm{Var}X \leqslant m/k$ and so Chebyshev's inequality implies that

$$\mathbb{P}[X \leqslant (k-1)m/k - \sqrt{2\ell m/k}] \leqslant \mathbb{P}[|X - \mathbb{E}X| \geqslant \sqrt{2\ell m/k}] \leqslant \frac{k\cdot \mathrm{Var}X}{2\ell m} \leqslant \frac{1}{2\ell}.$$

Theorem 1.2 now follows by summing up this bound on the failure probability for each of the graphs G_i. □

Proof of Theorem 1.3. Let ε be as in the statement of the theorem. As in the previous proof, we first consider a single graph G, this time with m edges and maximum degree $\Delta \leqslant \varepsilon m$. Consider a random partition of $V := V(G)$ into k disjoint sets V_j which is obtained by including each vertex $v \in V$ into V_j with probability $1/k$ independently of all other vertices. Fix some s and t with $1 \leqslant s < t \leqslant k$. This time let $X_e := 1$ if one endvertex of e is contained in X_s and the other in X_t. Put $X_e := 0$ otherwise. So $\mathbb{P}[X_e = 1] = 2/k^2 =: \alpha$. Now the value of $\mathbb{E}[X_e \cdot X_{e'}]$ depends on whether or not e and e' have an endvertex in common. If they do have an endvertex in common, we will use the trivial bound $\mathbb{E}[X_e \cdot X_{e'}] \leqslant 1 < 1 + \alpha^2$. Note that the number of ordered pairs e, e' of distinct edges for which this can happen is trivially at most $2\Delta m$. If e and e' have no vertex in common, then it is easy to see that

$$\mathbb{E}[X_e \cdot X_{e'}] = \mathbb{P}[X_e = 1]\mathbb{P}[X_{e'} = 1] = \alpha^2.$$

Let $X := \sum_{e\in E(G)} X_e$. Thus $\mathbb{E}[X] = 2m/k^2 = \alpha m$. Moreover,

$$\mathbb{E}[X^2] = \sum_{e\in E(G)} \mathbb{E}[X_e] + \sum_{\substack{e,e'\in E(G)\\ e\neq e'}} \mathbb{E}[X_e \cdot X_{e'}]$$
$$< \mathbb{E}[X] + 2\Delta m + \sum_{\substack{e,e'\in E(G)\\ e\neq e'}} \alpha^2$$
$$\leqslant \alpha m + 2\Delta m + \alpha^2 m^2 \leqslant 3\Delta m + (\mathbb{E}[X])^2.$$

Thus $\mathrm{Var}X \leqslant 3\Delta m \leqslant 3\varepsilon m^2$. So we can conclude that

$$\mathbb{P}[X \leqslant \alpha m - \varepsilon^{1/4}m] \leqslant \mathbb{P}[|X - \mathbb{E}X| \geqslant \varepsilon^{1/4}m] \leqslant \frac{\mathrm{Var}X}{\sqrt{\varepsilon}m^2} \leqslant 3\sqrt{\varepsilon} \leqslant \frac{1}{\ell k^2}.$$

In exactly the same way one can show that $\mathbb{P}[e_G(V_s) \leqslant m/k^2 - \varepsilon^{1/4}m] \leqslant 1/(\ell k^2)$. (This time $\alpha := 1/k^2$.) Now sum up these failure probabilities for all the $\binom{k}{2}$ pairs s, t and all the k values of s to see that the probability that a random partition does not have the required properties for G is at most $3/(4\ell)$. Again, Theorem 1.3 follows from summing up this probability for all G_i. □

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We remark that at the expense of increasing the error terms the partition classes in Theorems 1.1–1.3 can be chosen to have almost equal sizes. Indeed, Chernoff's inequality implies that in a random partition of the vertex set as considered in the proofs with high probability the vertex classes have almost equal sizes.

4. Algorithmic aspects

Papadimitriou and Yannakakis [11] showed that the Max-Cut problem is APX-complete. On the other hand, as mentioned in the Introduction, the obvious greedy algorithm always guarantees a cut whose size is at least $m/2$. Moreover, the proofs described in the previous section can be derandomized to yield polynomial algorithms which construct partitions satisfying the bounds in Theorems 1.1–1.3. As the derandomization argument is similar for all three results, we only describe it for Theorem 1.1. More background information on derandomization can be found, for instance, in the books [1, 10] and in Fundia [8] (in particular, the framework described in the latter applies to our situation). For simplicity, we consider Theorem 1.1 only for $\ell = 2$, *i.e.*, in the case of two graphs.

So let G_1 and G_2 be two graphs whose vertex set is V with $e(G_i) = m_i$. Consider a random partition of V into sets A and B as described in the proof of Theorem 1.1 (*cf.* Lemma 3.1). For $i = 1, 2$ define random variables $X_i := e_{G_i}(A, B)$ and put $\mu_i := m_i/2 = \mathbb{E}[X_i]$. Set

$$Z_i := \frac{\mu_i^2 - 2\mu_i X_i + X_i^2}{m_i}$$

for $i = 1, 2$ and $Z := Z_1 + Z_2$. The proof of Theorem 1.1 shows that, for each i,

$$\mathbb{P}[X_i < \mu_i - \sqrt{m_i}] \leqslant \frac{\mathrm{Var}X_i}{m_i} < 1/2.$$

But $\mathbb{E}[Z_i] = \mathrm{Var}X_i/m_i$ and so $\mathbb{E}[Z] = \mathbb{E}[Z_1] + \mathbb{E}[Z_2] < 1$. Let $v_1, \ldots, v_n$ be an enumeration of the vertices in V. Let A_i denote the event that the vertex v_i is contained in A. Then

$$1 > \mathbb{E}[Z] = (\mathbb{E}[Z \mid A_1] + \mathbb{E}[Z \mid A_1^c])/2 \geqslant \min\{\mathbb{E}[Z \mid A_1], \mathbb{E}[Z \mid A_1^c]\}.$$

Thus at least one of $\mathbb{E}[Z \mid A_1]$, $\mathbb{E}[Z \mid A_1^c]$ has to be less than 1. Let $C_1 \in \{A_1, A_1^c\}$ be such that $\mathbb{E}[Z \mid C_1] < 1$. Note that both $\mathbb{E}[Z \mid A_1]$ and $\mathbb{E}[Z \mid A_1^c]$ can be computed in polynomial time and so C_1 can also be determined in polynomial time. Now

$$1 > \mathbb{E}[Z \mid C_1] = (\mathbb{E}[Z \mid C_1 \cap A_2] + \mathbb{E}[Z \mid C_1 \cap A_2^c])/2.$$

So, as before, there exists $C_2 \in \{A_2, A_2^c\}$ such that $\mathbb{E}[Z \mid C_1 \cap C_2] < 1$ and C_2 can be determined in polynomial time. We continue in this fashion until we have obtained events $C_k \in \{A_k, A_k^c\}$ for all $k = 1, \ldots, n$ such that

$$\mathbb{E}[Z \mid C_1 \cap \cdots \cap C_n] < 1.$$

The proof of Chebyshev's inequality shows that, for each $i = 1, 2$ and for any event U which has positive probability, we have

$$\mathbb{P}[X_i < \mu_i - \sqrt{m_i} \mid U] \leqslant \frac{\mu_i^2 - 2\mu_i \mathbb{E}[X_i \mid U] + \mathbb{E}[X_i^2 \mid U]}{m_i} = \mathbb{E}[Z_i \mid U]$$

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(the above also follows from Corollary 4 in [8]). Taking $U := C_1 \cap \cdots \cap C_n$, this implies that

$$\sum_{i=1,2} \mathbb{P}[X_i < \mu_i - \sqrt{m_i} \mid U] \leqslant \sum_{i=1,2} \mathbb{E}[Z_i \mid U] = \mathbb{E}[Z \mid U] < 1. \tag{4.1}$$

But $U := C_1 \cap \cdots \cap C_n$ means that for each vertex $v_k \in V$ we have decided whether $v_k \in A$ or $v_k \in B$. So the left-hand side of (4.1) is either 0 or 1, *i.e.*, it has to be 0. This means that the unique partition corresponding to $C_1 \cap \cdots \cap C_n$ is as desired in Theorem 1.1. Since each C_k can be determined in polynomial time, this gives us a polynomial algorithm.

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References

[1] Alon, N. and Spencer, J. (2000) *The Probabilistic Method*, 2nd edn, Wiley-Interscience.
[2] Bollobás, B. and Scott, A. D. (1993) Judicious partitions of graphs. *Periodica Math. Hungar.* **26** 127–139.
[3] Bollobás, B. and Scott, A. D. (1999) Exact bounds for judicious partitions. *Combinatorica* **19** 473–486.
[4] Bollobás, B. and Scott, A. D. (2002) Problems and results on judicious partitions. *Random Struct. Alg.* **21** 414–430.
[5] Bollobás, B. and Scott, A. D. (2004) Judicious partitions of bounded-degree graphs. *J. Graph Theory* **46** 131–143.
[6] Edwards, C. S. (1973) Some extremal properties of bipartite subgraphs. *Canadian J. Math.* **25** 475–485.
[7] Edwards, C. S. (1975) An improved lower bound on the number of edges in a largest bipartite subgraph. In *Proc. 2nd Czech. Symposium on Graph Theory*, Prague, pp. 167–181.
[8] Fundia, A. D. (1996) Derandomizing Chebyshev's inequality to find independent sets in uncrowded hypergraphs. *Random Struct. Alg.* **8** 131–147.
[9] Kühn, D. and Osthus, D. (2003) Partitions of graphs with high minimum degree or connectivity. *J. Combin. Ser. Theory B* **88** 29–43.
[10] Motwani, R. and Raghavan, P. (1995) *Randomized Algorithms*, Cambridge University Press.
[11] Papadimitriou, C. H. and Yannakakis, M. (1991) Optimization, approximation, and complexity classes. *J. Comput. System Sci.* **43** 425–440.
[12] Porter, T. D. (1992) On a bottleneck conjecture of Erdős. *Combinatorica* **12** 317–321.
[13] Porter, T. D. (1994) Graph partitions. *J. Combin. Math. Combin. Comp.* **15** 111–118.
[14] Porter, T. D. (1999) Minimal partitions of a graph. *Ars Combinatorica* **53** 181–186.
[15] Rautenbach, D. and Szigeti, Z. (2004) Simultaneous large cuts. Manuscript.
[16] Stiebitz, M. (1996) Decomposing graphs under degree constraints. *J. Graph Theory* **23** 321–324.

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