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**Citation for published version (APA):**

Aronov, B., de Berg, M. T., Gudmundsson, J., & Horton, M. (2020). On beta-Plurality Points in Spatial Voting Games. *arXiv*, 2020, [2003.07513]. <https://doi.org/10.48550/arXiv.2003.07513>

**DOI:**

[10.48550/arXiv.2003.07513](https://doi.org/10.48550/arXiv.2003.07513)

**Document status and date:**

Published: 17/03/2020

**Document Version:**

Publisher's PDF, also known as Version of Record (includes final page, issue and volume numbers)

**Please check the document version of this publication:**

- A submitted manuscript is the version of the article upon submission and before peer-review. There can be important differences between the submitted version and the official published version of record. People interested in the research are advised to contact the author for the final version of the publication, or visit the DOI to the publisher's website.
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- The final published version features the final layout of the paper including the volume, issue and page numbers.

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# On $\beta$ -Plurality Points in Spatial Voting Games

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## Abstract

Let  $V$  be a set of  $n$  points in  $\mathbb{R}^d$ , called *voters*. A point  $p \in \mathbb{R}^d$  is a *plurality point* for  $V$  when the following holds: for every  $q \in \mathbb{R}^d$  the number of voters closer to  $p$  than to  $q$  is at least the number of voters closer to  $q$  than to  $p$ . Thus, in a vote where each  $v \in V$  votes for the nearest proposal (and voters for which the proposals are at equal distance abstain), proposal  $p$  will not lose against any alternative proposal  $q$ . For most voter sets a plurality point does not exist. We therefore introduce the concept of  $\beta$ -*plurality points*, which are defined similarly to regular plurality points except that the distance of each voter to  $p$  (but not to  $q$ ) is scaled by a factor  $\beta$ , for some constant  $0 < \beta \leq 1$ . We investigate the existence and computation of  $\beta$ -plurality points, and obtain the following results.

- Define  $\beta_d^* := \sup\{\beta : \text{any finite multiset } V \text{ in } \mathbb{R}^d \text{ admits a } \beta\text{-plurality point}\}$ . We prove that  $\beta_2^* = \sqrt{3}/2$ , and that  $1/\sqrt{d} \leq \beta_d^* \leq \sqrt{3}/2$  for all  $d \geq 3$ .
- Define  $\beta(V) := \sup\{\beta : V \text{ admits a } \beta\text{-plurality point}\}$ . We present an algorithm that, given a voter set  $V$  in  $\mathbb{R}^d$ , computes an  $(1 - \varepsilon) \cdot \beta(V)$  plurality point in time  $O(\frac{n^2}{\varepsilon^{3d-2}} \cdot \log \frac{n}{\varepsilon^{d-1}} \cdot \log^2 \frac{1}{\varepsilon})$ .

**2012 ACM Subject Classification** General and reference  $\rightarrow$  General literature; Theory of computation  $\rightarrow$  Design and analysis of algorithms

**Keywords and phrases** Computational geometry, Spatial voting theory, Plurality point, Computational social choice

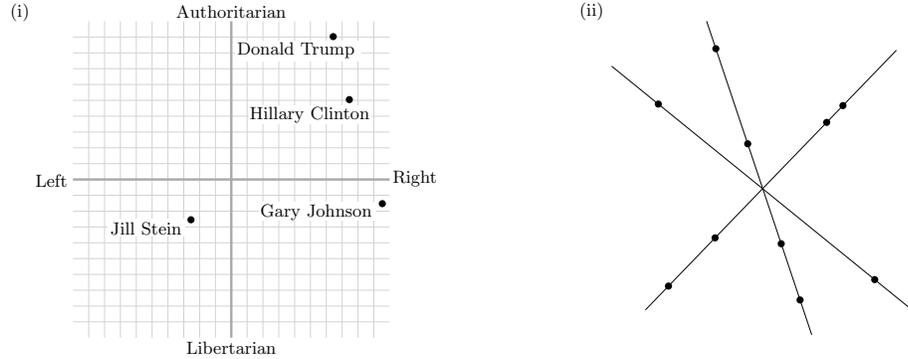
**Funding** *Boris Aronov*: Partially supported by NSF grant CCF-15-40656 and by grant 2014/170 from the US-Israel Binational Science Foundation.

*Mark de Berg*: Supported by the Netherlands' Organisation for Scientific Research (NWO) under project no. 024.002.003.

*Joachim Gudmundsson*: Supported under the Australian Research Council Discovery Projects funding scheme (project numbers DP150101134 and DP180102870).

## 1 Introduction

**Background.** Voting theory is concerned with mechanisms to combine preferences of individual voters into a collective decision. A desirable property of such a collective decision is that it is stable, in the sense that no alternative is preferred by more voters. In spatial voting games [5, 10] this is formalized as follows; see Fig. 1(i) for an example in a political context. The space of all possible decisions is modeled as  $\mathbb{R}^d$  and every voter is represented by a point in  $\mathbb{R}^d$ , where the dimensions represent different aspects of the decision and the point representing a voter corresponds to the ideal decision for that voter. A voter  $v$  now prefers a proposed decision  $p \in \mathbb{R}^d$  over some alternative proposal  $q \in \mathbb{R}^d$  when  $v$  is closer



■ **Figure 1** (i) The US presidential candidates 2016 modelled in the spatial voting model, according to The Political Compass (<https://politicalcompass.org/uselection2016>). Note that the points representing voters are not shown. (ii) The point set satisfies the generalized Plott symmetry conditions and therefore admits a plurality point.

to  $p$  than to  $q$ . Thus a point  $p \in \mathbb{R}^d$  represents a stable decision for a given finite set  $V$  of voters if, for any alternative  $q \in \mathbb{R}^d$ , we have  $|\{v \in V : |vp| < |vq|\}| \geq |\{v \in V : |vq| < |vp|\}|$ . Such a point  $p$  is called a *plurality point*.<sup>1</sup>

For  $d = 1$ , a plurality point always exists, since in  $\mathbb{R}^1$  a median of  $V$  is a plurality point. This is not true in higher dimensions, however. Define a *median hyperplane* for a set  $V$  of voters to be a hyperplane  $h$  such that both open half-spaces defined by  $h$  contain fewer than  $|V|/2$  voters. For  $d \geq 2$  a plurality point in  $\mathbb{R}^d$  exists if and only if all median hyperplanes for  $V$  meet in a common point; see Fig. 1(ii). This condition is known as *generalized Plott symmetry* conditions [12, 23]; see also the papers by Wu *et al.* [28] and de Berg *et al.* [4], who present algorithms to determine the existence of a plurality point for a given set of voters.

It is very unlikely that voters are distributed in such a way that all median hyperplanes have a common intersection. (Indeed, if this happens, then a slightest generic perturbation of a single voter destroys the existence of the plurality point.) When a plurality point does not exist, we may want to find a point that is close to being a plurality point. One way to formalize this is to consider the center of the *yolk* (or *plurality ball*) of  $V$ , where the yolk [14, 17, 20, 22] is the smallest ball intersecting every median hyperplane of  $V$ . We introduce  $\beta$ -plurality points as an alternative way to relax the requirements for a plurality point, and study several combinatorial and algorithmic questions regarding  $\beta$ -plurality points.

**$\beta$ -Plurality points: definition and main questions.** Let  $V$  be a multiset<sup>2</sup> of  $n$  voters in  $\mathbb{R}^d$  in arbitrary, possibly coinciding, positions. In the traditional setting a proposed point  $p \in \mathbb{R}^d$  wins a voter  $v \in V$  against an alternative  $q$  if  $|pv| < |qv|$ . We relax this by fixing a parameter  $\beta$  with  $0 < \beta \leq 1$  and letting  $p$  win  $v$  against  $q$  if  $\beta \cdot |pv| < |qv|$ . Thus we give an advantage to the initial proposal  $p$  by scaling distances to  $p$  by a factor  $\beta \leq 1$ . We now define

$$V[p \succ_{\beta} q] := \{v \in V : \beta \cdot |pv| < |qv|\} \quad \text{and} \quad V[p \prec_{\beta} q] := \{v \in V : \beta \cdot |pv| > |qv|\}$$

<sup>1</sup> One can also require  $p$  to be strictly more popular than any alternative  $q$ . This is sometimes called a *strong* plurality point, in contrast to the *weak* plurality points that we consider.

<sup>2</sup> Even though we allow  $V$  to be a multiset, we sometimes refer to it as a “set” to ease the reading. When the fact that  $V$  is a multiset requires special treatment, we explicitly address this.

to be the multisets of voters won by  $p$  over  $q$  and lost by  $p$  against  $q$ , respectively. Finally, we say that a point  $p \in \mathbb{R}^d$  is a  $\beta$ -plurality point for  $V$  when

$$|V[p \succ_{\beta} q]| \geq |V[p \prec_{\beta} q]|, \quad \text{for any point } q \in \mathbb{R}^d.$$

Observe that  $\beta$ -plurality is *monotone* in the sense that if  $p$  is a  $\beta$ -plurality point then  $p$  is also a  $\beta'$ -plurality point for all  $\beta' < \beta$ .

The spatial voting model was popularised by Black [5] and Down [10] in the 1950s. Stokes [26] criticized its simplicity and was the first to highlight the importance of taking non-spatial aspects into consideration. The reasoning is that voters may evaluate a candidate not only on their policies—their position in the policy space—but also take their so-called *valence* into account: charisma, competence, or other desirable qualities in the public’s mind [13]. A candidate can also increase her valence by a stronger party support [27] or campaign spending [18]. Several models have been proposed to bring the spatial model closer to a more realistic voting approach; see [15, 16, 24] as examples. A common model is the multiplicative model, introduced by Hollard and Rossignol [19], which is closely related to the concept of a  $\beta$ -plurality point. The multiplicative model augments the existing spatial utility function by scaling the candidate’s valence by a multiplicative factor. Note that in the 2-player game considered in this paper the multiplicative model is the same as our  $\beta$ -plurality model. From a computational point of view very little is known about the multiplicative model. We are only aware of a result by Chung [7], who studied the problem of positioning a new candidate in an existing space of voters and candidates, so that the valence required to win at least a given number of voters is minimized.

One reason for introducing  $\beta$ -plurality was that a set  $V$  of voters in  $\mathbb{R}^d$ , for  $d \geq 2$ , generally does not admit a plurality point. This immediately raises the question: Is it true that, for  $\beta$  small enough, any set  $V$  admits a  $\beta$ -plurality point? If so, we want to know the largest  $\beta$  such that any voter set  $V$  admits a  $\beta$ -plurality point, that is, we wish to determine

$$\beta_d^* := \sup\{\beta : \text{any finite multiset } V \text{ in } \mathbb{R}^d \text{ admits a } \beta\text{-plurality point}\}.$$

Note that  $\beta_1^* = 1$ , since any set  $V$  in  $\mathbb{R}^1$  admits a plurality point and 1-plurality is equivalent to the traditional notion of plurality.

After studying this combinatorial problem in Section 2, we turn our attention to the following algorithmic question: given a voter set  $V$ , find a point  $p$  that is a  $\beta$ -plurality point for the largest possible value  $\beta$ . In other words, if we define

$$\beta(V) := \sup\{\beta : V \text{ admits a } \beta\text{-plurality point}\}$$

and

$$\beta(p, V) := \sup\{\beta : p \text{ is a } \beta\text{-plurality point for } V\}$$

then we want to find a point  $p$  such that  $\beta(p, V) = \beta(V)$ .

**Outline.** In Section 2 we prove that  $\beta_d^* \leq \sqrt{3}/2$  for all  $d \geq 2$ . To this end we first show that  $\beta_d^*$  is non-increasing in  $d$ , and then we exhibit a voter set  $V$  in  $\mathbb{R}^2$  such that  $\beta(V) \leq \sqrt{3}/2$ . We also show how to construct, for any given  $V$  in  $\mathbb{R}^2$ , a  $(\sqrt{3}/2)$ -plurality point, thus proving that  $\beta_2^* = \sqrt{3}/2$ . For  $d \geq 3$  we show how to construct a  $(1/\sqrt{d})$ -plurality point.

In Section 3 we study the problem of computing, for a given voter set  $V$  of  $n$  points in  $\mathbb{R}^d$ , a  $\beta$ -plurality point for the largest possible  $\beta$ . (Here we assume  $d$  to be a fixed constant.)

While such a point can be found in polynomial time, the resulting running time is quite high. We therefore focus our attention on finding an approximately optimal point  $p$ , that is, a point  $p$  such that  $\beta(p, V) \geq (1 - \varepsilon) \cdot \beta(V)$ . We show that such a point can be computed in  $O(\frac{n^2}{\varepsilon^{3d-2}} \cdot \log \frac{n}{\varepsilon^{d-1}} \cdot \log^2 \frac{1}{\varepsilon})$  time.

**Notation.** We denote the *open* ball of radius  $\rho$  centered at a point  $q \in \mathbb{R}^d$  by  $B(q, \rho)$  and, for a point  $p \in \mathbb{R}^d$  and a voter  $v$ , we define  $D_\beta(p, v) := B(v, \beta \cdot |pv|)$ . Observe that  $p$  wins  $v$  against a competitor  $q$  if and only if  $q$  is strictly outside  $D_\beta(p, v)$ , while  $q$  wins  $v$  if and only if  $q$  is strictly inside  $D_\beta(p, v)$ . Hence,  $V[p \prec_\beta q] = \{v \in V : q \in D_\beta(p, v)\}$ . We define  $\mathcal{D}_\beta(p) := \{D_\beta(p, v) : v \in V\}$ —here we assume  $V$  is clear from the context—and let  $\mathcal{A}(\mathcal{D}_\beta(p))$  denote the arrangement induced by  $\mathcal{D}_\beta(p)$ . The competitor point  $q$  that wins the most voters against  $p$  will thus lie in the cell of  $\mathcal{A}(\mathcal{D}_\beta(p))$  of the greatest depth or, more precisely, the cell contained in the maximum number of disks  $D_\beta(p, v)$ .

## 2 Bounds on $\beta_d^*$

In this section we will prove bounds on  $\beta_d^*$ , the supremum of all  $\beta$  such that any finite set  $V \subset \mathbb{R}^d$  admits a  $\beta$ -plurality point. We start with an observation that allows us to apply bounds on  $\beta_d^*$  to those on  $\beta_{d'}^*$  for  $d' > d$ . Let  $\text{conv}(V)$  denote the convex hull of  $V$ .

- **Observation 2.1.** *Let  $V$  be a finite multiset of voters in  $\mathbb{R}^d$ .*
- (i) *Suppose a point  $p \in \mathbb{R}^d$  is not a  $\beta$ -plurality point for  $V$ . Then there is a point  $q \in \text{conv}(V)$  such that  $|V[p \succ_\beta q]| < |V[p \prec_\beta q]|$ .*
  - (ii) *For any  $p' \notin \text{conv}(V)$ , there is a point  $p \in \text{conv}(V)$  with  $\beta(p, V) > \beta(p', V)$ .*
  - (iii) *For any  $d' > d$  we have  $\beta_{d'}^* \leq \beta_d^*$ .*

**Proof.** Note that for every point  $r \notin \text{conv}(V)$  there is a point  $r' \in \text{conv}(V)$  that lies strictly closer to all voters in  $V$ , namely the point  $r' \in \partial \text{conv}(V)$  closest to  $r'$ . This immediately implies part (i): if  $p$  is beaten by some point  $q \notin \text{conv}(V)$  then  $p$  is certainly beaten by a point  $q' \in \text{conv}(V)$  that lies strictly closer to all voters in  $V$  than  $q$ . It also immediately implies part (ii), because if a point  $p$  lies strictly closer to all voters in  $V$  than a point  $p'$ , then  $\beta(p, V) > \beta(p', V)$ .

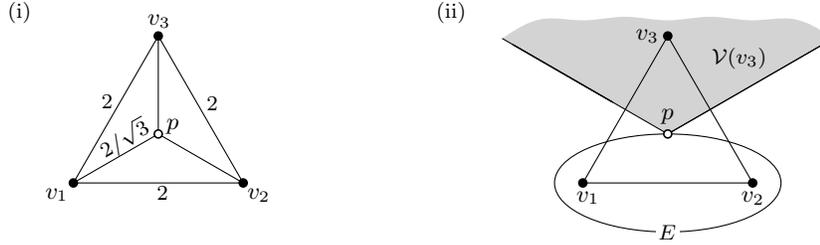
To prove part (iii), let  $V \in \mathbb{R}^d$  be a voter set such that  $\beta(V) = \beta_d^*$ . Now embed  $V$  into  $\mathbb{R}^{d'}$ , say in the flat  $x_{d+1} = \dots = x_{d'} = 0$ , obtaining a set  $V'$ . Then  $\beta(V') = \beta(V)$  by parts (i) and (ii). Hence,  $\beta_{d'}^* \leq \beta(V') = \beta(V) = \beta_d^*$ . ◀

We can now prove an upper bound on  $\beta_d^*$ .

- **Lemma 2.2.**  $\beta_d^* \leq \sqrt{3}/2$ , for  $d \geq 2$ .

**Proof.** By Observation 2.1(iii), it suffices to prove the lemma for  $d = 2$ . To this end let  $V = \{v_1, v_2, v_3\}$  consist of three voters that form an equilateral triangle  $\Delta$  of side length 2 in  $\mathbb{R}^2$ ; see Fig. 2(i).

Let  $p$  denote the center of  $\Delta$ . We will first argue that  $\beta(p, V) = \sqrt{3}/2$ . Note that  $|pv_i| = 2/\sqrt{3}$  for all three voters  $v_i$ . Hence, for  $\beta = \sqrt{3}/2$ , the open balls  $D_\beta(v_i, p)$  are pairwise disjoint and touching at the mid-points of the edges of  $\Delta$ . Therefore any competitor  $q$  either wins one voter and loses the remaining two, or wins no voter and loses at least one. The former happens when  $q$  lies inside one of the three balls  $D_\beta(v_i, p)$ ; the later happens when  $q$  does not lie inside any of the balls, because in that case  $q$  can be on the boundary of at most two of the balls. Thus, for  $\beta = \sqrt{3}/2$ , the point  $p$  always wins more voters than  $q$



■ **Figure 2** (i) The set  $V = \{v_1, v_2, v_3\}$  of voters and the point  $p$  used in the proof of Lemma 2.2. (ii) The ellipse  $E$  is tangent to the Voronoi cell  $\mathcal{V}(v_3)$ .

does. On the other hand, for  $\beta > \sqrt{3}/2$ , any two balls  $D_\beta(v_i, p)$ ,  $D_\beta(v_j, p)$  intersect and so a point  $q$  located in such a pairwise intersection wins two voters and beats  $p$ . We conclude that  $\beta(p, V) = \sqrt{3}/2$ , as claimed.

The lemma now follows if we can show that  $\beta(p', V) \leq \sqrt{3}/2$  for any  $p' \neq p$ . Let  $\text{Vor}(V)$  be the Voronoi diagram of  $V$ , and let  $\mathcal{V}(v_i)$  be the closed Voronoi cell of  $v_i$ , as shown in Fig. 2(ii). Assume without loss of generality that  $p'$  lies in  $\mathcal{V}(v_3)$ . Let  $E$  be the ellipse with foci  $v_1$  and  $v_2$  that passes through  $p$ . Thus

$$E := \{z \in \mathbb{R}^2 : |zv_1| + |zv_2| = 4/\sqrt{3}\}.$$

Note that  $E$  is tangent to  $\mathcal{V}(v_3)$  at the point  $p$ . Hence, any point  $p' \neq p$  in  $\mathcal{V}(v_3)$  has  $|p'v_1| + |p'v_2| > 4/\sqrt{3}$ . This implies that for  $\beta \geq \sqrt{3}/2$  we have  $\beta \cdot |p'v_1| + \beta \cdot |p'v_2| > 2$ , and so the disks  $D_\beta(p', v_1)$  and  $D_\beta(p', v_2)$  intersect. It follows that for  $\beta \geq \sqrt{3}/2$  there is a competitor  $q$  that wins two voters against  $p'$ , which implies  $\beta(p', V) \leq \sqrt{3}/2$  and thus finishes the proof of the lemma. ◀

We now prove lower bounds on  $\beta_d^*$ . We first prove that  $\beta_d^* \geq 1/\sqrt{d}$  for any  $d \geq 2$ , and then we improve the lower bound to  $\sqrt{3}/2$  for  $d = 2$ . The latter bound is tight by Lemma 2.2.

Let  $V$  be a finite multiset of  $n$  voters in  $\mathbb{R}^d$ . We call a hyperplane  $h$  *balanced* with respect to  $V$ , if both open half-spaces defined by  $h$  contain at most  $n/2$  voters from  $V$ . Note the difference with median hyperplanes, which are required to have fewer than  $n/2$  voters in both open half-spaces. Clearly, for any  $1 \leq i \leq d$  there is a balanced hyperplane orthogonal to the  $x_i$ -axis, namely the hyperplane  $x_i = m_i$ , where  $m_i$  is a median in the multiset of all  $x_i$ -coordinates of the voters in  $V$ . (In fact, for any direction  $\vec{d}$  there is a balanced hyperplane orthogonal to  $\vec{d}$ .)

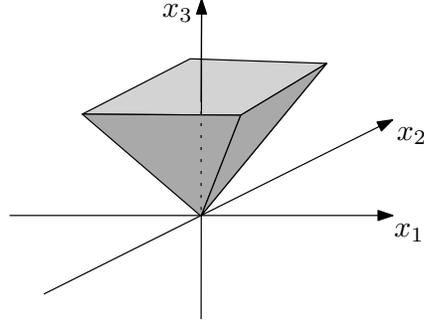
► **Lemma 2.3.** *Let  $d \geq 2$ . For any finite multi-set  $V$  of voters in  $\mathbb{R}^d$  there exists a point  $p \in \mathbb{R}^d$  such that  $\beta(p, V) = 1/\sqrt{d}$ . Moreover, such a point  $p$  can be computed in  $O(n)$  time.*

**Proof.** Let  $\mathcal{H} := \{h_1, \dots, h_d\}$  be a set of balanced hyperplanes with respect to  $V$  such that  $h_i$  is orthogonal to the  $x_i$ -axis, and assume without loss of generality that  $h_i: x_i = 0$ . We will prove that the point  $p$  located at the origin is a  $\beta$ -plurality point for  $V$  for any  $\beta < 1/\sqrt{d}$ , thus showing that  $\beta(p, V) \geq 1/\sqrt{d}$ .

Let  $q = (q_1, \dots, q_d)$  be any competitor of  $p$ . We can assume without loss of generality that  $\max_{1 \leq i \leq d} |q_i| = q_d > 0$ . Thus  $q$  lies in the closed cone  $C_d^+$  defined as

$$C_d^+ := \{ (x_1, \dots, x_d) \in \mathbb{R}^d : x_d \geq |x_j| \text{ for all } j \neq d \}.$$

Note that  $C_d^+$  is bounded by portions of the  $2(d-1)$  hyperplanes  $x_d = \pm x_j$  with  $j \neq d$ ; see Fig. 3.



■ **Figure 3** The cone  $C_3^+$  used in the proof of Lemma 2.3.

Because  $h_d: x_d = 0$  is a balanced hyperplane, the open halfspace  $h_d^+: x_d > 0$  contains at most  $n/2$  voters, which implies that the closed halfspace  $\text{cl}(h_d^-): x_d \leq 0$  contains at least  $n/2$  voters. Hence, it suffices to argue that for any  $\beta < 1/\sqrt{d}$  the point  $p$  wins all the voters in  $\text{cl}(h_d^-)$  against  $q$ .

*Claim.* For any voter  $v \in \text{cl}(h_d^-)$  with  $v \neq p$ , we have that  $\sin(\angle qpv) \geq 1/\sqrt{d}$  with equality if and only if  $q$  lies on an edge of  $C_d^+$  and  $v$  lies on the orthogonal projection of this edge onto  $h_d$ .

*Proof.* For any point  $v$  below  $h_d$  there is a point  $v' \in h_d$  with  $\angle qpv' < \angle qpv$ , namely the orthogonal projection of  $v$  onto  $h_d$ . Hence, from now on we assume that  $v \in h_d$ . First, we prove that  $\sin(\angle qpv) = 1/\sqrt{d}$  if  $q$  lies on an edge  $e$  of  $C_d^+$  and  $v$  lies on the orthogonal projection  $\bar{e}$  of  $e$  onto  $h_d$ . Assume without loss of generality that  $e$  is the edge of  $C_d^+$  defined by the intersection of the  $d - 1$  hyperplanes  $x_d = x_j$ , so that  $q_1 = \dots = q_{d-1} = q_d$ . Since  $\angle qpv$  is the same for any  $v \in \bar{e}$ , we may assume that  $v$  is the orthogonal projection of  $q$  to  $h_d$ , which means  $|qv| = q_d$ . We then have

$$\sin(\angle qpv) = \frac{|qv|}{|pq|} = \frac{q_d}{\sqrt{q_1^2 + \dots + q_d^2}} = \frac{1}{\sqrt{d}}.$$

Now assume the condition for equality does not hold. Let  $\rho$  be the ray starting at  $p$  and containing  $q$ , and let  $\bar{\rho}$  be its orthogonal projection onto  $h_d$ . We have two cases:  $v \in \bar{\rho}$  but  $q$  is not contained in an edge of  $C_d^+$ , or  $v \notin \bar{\rho}$ .

In the former case we may, as before, assume that  $v$  is the projection of  $q$  onto  $h_d$ . Since  $q \in C_d^+$  we have  $q_d \geq |q_j|$  for all  $j$ . Moreover, since  $q$  does not lie on an edge of  $C_d^+$  we have  $q_d > |q_{j^*}|$  for at least one  $j^*$ . Thus  $|pq| = \sqrt{q_1^2 + \dots + q_d^2} < \sqrt{d \cdot q_d^2} = \sqrt{d} \cdot |qv|$ , and  $\sin(\angle qpv) = |qv|/|pq| > 1/\sqrt{d}$ .

In the latter case, let  $\ell$  be the line containing  $p$  and  $v$ , and let  $v'$  be the point on  $\ell$  closest to  $q$ . Then  $|qv| \geq |qv'| > |q\bar{q}|$ , where  $\bar{q}$  is the projection of  $q$  onto  $h_d$ , and so

$$\sin(\angle qpv) \geq \frac{|qv'|}{|pq|} > \frac{|q\bar{q}|}{|pq|} \geq \frac{1}{\sqrt{d}}. \quad \triangleleft$$

We can now use the Law of Sines and the claim above to derive that for any  $\beta < 1/\sqrt{d}$  and any voter  $v \in \text{cl}(h_d^-)$  with  $v \neq p$  we have

$$\beta \cdot |pv| < \frac{1}{\sqrt{d}} \cdot |pv| = \frac{1}{\sqrt{d}} \cdot \frac{|qv| \cdot \sin(\angle pqv)}{\sin(\angle qpv)} \leq |qv| \cdot \sin(\angle pqv) \leq |qv|.$$



■ **Figure 4** (i) The balanced line  $\mu(\theta)$ . (ii) If  $p_{23}$  is to the left of the directed line  $\ell_1(0)$  then  $p_{13}(0)$  is to the right of  $\ell_2(0)$ .

Hence,  $p$  wins every point in  $\text{cl}(h_d^-)$ . This proves the first part of the lemma since  $\text{cl}(h_d^-)$  contains at least  $n/2$  voters, as already remarked.

Computing the point  $p$  is trivial once we have the balanced hyperplanes  $h_i$ , which can be found in  $O(n)$  time by computing a median  $x_i$ -coordinate for each  $1 \leq i \leq d$ .

In  $\mathbb{R}^2$  we can improve the above bound: for any voter set  $V$  in the plane we can find a point  $p$  such that  $\beta(p, V) = \sqrt{3}/2$ . By Lemma 2.2 this bound is tight. The improvement is based on Lemma 2.4 below. This lemma—in fact a stronger version, stating that any two opposite cones defined by the three concurrent lines contain the same number of points—has been proved for even  $n$  by Dumitrescu *et al.* [11]. Our proof of Lemma 2.4 is similar to their proof. We give it because we also need it for odd  $n$ , and because we will need an understanding of the proof to describe our algorithm for computing the concurrent triple in the lemma. Our algorithm will run in  $O(n \log^2 n)$  time, a significant improvement over the  $O(n^{4/3} \log^{1+\varepsilon} n)$  running time obtained (for the case of even  $n$ ) by Dumitrescu *et al.* [11].

► **Lemma 2.4.** *Given a multiset  $V$  of  $n$  voters in  $\mathbb{R}^2$ , there exists a triple of concurrent balanced lines  $(\ell_1, \ell_2, \ell_3)$  such that the smaller angle between any two of them is  $\frac{\pi}{3}$ .*

**Proof.** Define the *orientation* of a line to be the counterclockwise angle it makes with the positive  $y$ -axis. Recall that for any given orientation  $\theta$  there exists at least one balanced line with orientation  $\theta$ . When  $n$  is odd this line is unique: it passes through the median of the voter set  $V$  when  $V$  is projected orthogonally onto a line orthogonal to the lines of orientation  $\theta$ . In the rest of the proof it will be convenient to have a unique balanced line for any orientation  $\theta$ . To achieve this when  $n$  is even, we simply delete an arbitrary voter from  $V$ . (If there are other voters at the same location, these voters are not deleted.) This is allowed because when  $|V|$  is even, a balanced line for  $V \setminus \{v\}$  is also a balanced line for  $V$ .

Now let  $\mu$  be the function that maps an angle value  $\theta$  to the unique balanced line  $\mu(\theta)$ ; see Figure 4(i). Note that  $\mu$  is continuous for  $0 \leq \theta < \pi$ . Let  $\ell_1(\theta) := \mu(\theta)$ , and  $\ell_2(\theta) := \mu(\theta + \frac{\pi}{3})$ , and  $\ell_3(\theta) := \mu(\theta + \frac{2\pi}{3})$ . For  $i \neq j$ , let  $p_{ij}(\theta) := \ell_i(\theta) \cap \ell_j(\theta)$  be the intersection point between  $\ell_i(\theta)$  and  $\ell_j(\theta)$ . If  $p_{23}(0) \in \ell_1(0)$  then the lines  $\ell_1(0), \ell_2(0), \ell_3(0)$  are concurrent and we are done. Otherwise, consider the situation at  $\theta = 0$  and imagine  $\ell_1(0)$  and  $\ell_2(0)$  to be directed in the positive  $y$ -direction, as in Fig. 4(ii). Clearly, if  $p_{23}(0)$  is to the left of the directed line  $\ell_1(0)$  then  $p_{13}(0)$  is to the right of the directed line  $\ell_2(0)$ , and vice versa. Now increase  $\theta$  from 0 to  $\pi/3$ , and note that  $\ell_1(\pi/3) = \ell_2(0)$  and  $p_{23}(\pi/3) = p_{13}(0)$ . Hence,  $p_{23}(\theta)$  lies to a different side of the directed line  $\ell_1(\theta)$  for  $\theta = 0$  than it does for  $\theta = \pi/3$ . Since both  $\ell_1(\theta)$  and  $p_{23}(\theta)$  move continuously, this implies that for some  $\bar{\theta} \in (0, \pi/3)$  the point  $p_{23}(\bar{\theta})$  crosses the line  $\ell_1(\bar{\theta})$ , and so the lines  $\ell_1(\bar{\theta}), \ell_2(\bar{\theta}), \ell_3(\bar{\theta})$  are concurrent. ◀

Next we show how to efficiently compute a triple as in Lemma 2.4. We follow the definitions and notation from the proof of Lemma 2.4. We will assume that  $n$  is odd, which, as argued, is without loss of generality.

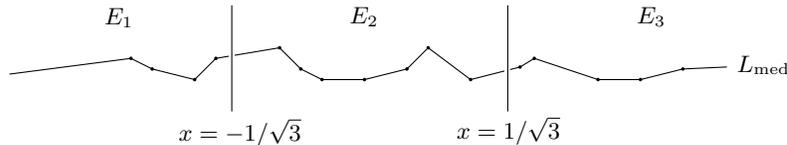
To find a concurrent triple of balanced lines, we first compute the lines  $\ell_1(0), \ell_2(0), \ell_3(0)$  in  $O(n)$  time. If they are concurrent, we are done. Otherwise, there is a  $\bar{\theta} \in (0, \pi/3)$  such that  $\ell_1(\bar{\theta}), \ell_2(\bar{\theta}), \ell_3(\bar{\theta})$  are concurrent. To find this value  $\bar{\theta}$ , we dualize the voter set  $V$ , using the standard duality transform that maps a point  $(a, b)$  to the line  $y = ax + b$ , and vice versa. Let  $v^*$  denote the dual line of the voter  $v$ , and let  $V^* := \{v^* : v \in V\}$ . Note that, for  $\theta \in (0, \pi/3)$ , the lines  $\ell_1(\theta), \ell_2(\theta), \ell_3(\theta)$  are all non-vertical, therefore their duals  $\ell_i^*(\theta)$  are well-defined.

Consider the arrangement  $\mathcal{A}(V^*)$  defined by the duals of the voters. For  $\theta \neq 0$ , define  $\text{slope}(\theta)$  to be the slope of the lines with orientation  $\theta$ . Then  $\mu^*(\theta)$ , the dual of  $\mu(\theta)$ , is the intersection point of the vertical line  $x = \text{slope}(\theta)$  with  $L_{\text{med}}$ , the median level in  $\mathcal{A}(V^*)$ . (The median level of  $\mathcal{A}(V^*)$  is the set of points  $q$  such that there are fewer than  $n/2$  lines below  $q$  and fewer than  $n/2$  lines above  $q$ ; this is well defined since we assume  $n$  is odd. The median level forms an  $x$ -monotone polygonal curve along edges of  $\mathcal{A}(V^*)$ .)

Now consider the duals  $\ell_1^*(\theta), \ell_2^*(\theta), \ell_3^*(\theta)$ , which all lie on  $L_{\text{med}}$ . For  $\theta \in (0, \pi/3)$ , the  $x$ -coordinate of  $\ell_1^*(\theta)$  lies in  $(-\infty, -1/\sqrt{3})$ , the  $x$ -coordinate of  $\ell_2^*(\theta)$  lies in  $(-1/\sqrt{3}, 1/\sqrt{3})$ , and the  $x$ -coordinate of  $\ell_3^*(\theta)$  lies in  $(1/\sqrt{3}, \infty)$ . We split  $L_{\text{med}}$  into three pieces corresponding to these ranges of  $x$ -coordinate. Let  $E_1, E_2$ , and  $E_3$  denote the sets of edges forming the parts of  $L_{\text{med}}$  in the first, second, and third range, respectively, where edges crossing the vertical lines  $x = -1/\sqrt{3}$  and  $x = 1/\sqrt{3}$  are split; see Fig. 5.

Recall that we want to find a value  $\bar{\theta} \in (0, \pi/3)$  such that  $\ell_1(\bar{\theta}), \ell_2(\bar{\theta}), \ell_3(\bar{\theta})$  are concurrent (or, in other words, such that the points  $\ell_1^*(\bar{\theta}), \ell_2^*(\bar{\theta}), \ell_3^*(\bar{\theta})$  are collinear). Also recall that, for any  $\theta \in (0, \pi/3)$ , the point  $\ell_i^*(\theta)$  lies on an edge in  $E_i$ , for  $i = 1, 2, 3$ . One way to find  $\bar{\theta}$  would be to explicitly compute  $L_{\text{med}}$ , and then increase  $\theta$  (starting at  $\theta = 0$ ) and see how the points  $\ell_i^*(\theta)$  move over  $E_i$ , until we reach a value where  $\ell_1(\bar{\theta}), \ell_2(\bar{\theta}), \ell_3(\bar{\theta})$  are concurrent. Since the best known bounds on the complexity of the median level is  $O(n^{4/3})$  [9] we will proceed differently, as follows.

1. Find an interval  $(\theta_1, \theta'_1) \subseteq (0, \pi/3)$  for which there is a  $\bar{\theta}$  with the desired properties and such that  $\ell_1^*(\theta)$  lies on the same edge of  $E_1$  for all  $\theta \in (\theta_1, \theta'_1)$ .
2. Find an interval  $(\theta_2, \theta'_2) \subseteq (\theta_1, \theta'_1)$  for which there is a  $\bar{\theta}$  with the desired properties and such that  $\ell_2^*(\theta)$  lies on the same edge of  $E_2$  for all  $\theta \in (\theta_2, \theta'_2)$ .
3. Find an interval  $(\theta_3, \theta'_3) \subseteq (\theta_2, \theta'_2)$  for which there is a  $\bar{\theta}$  with the desired properties and such that  $\ell_3^*(\theta)$  lies on the same edge of  $E_3$  for all  $\theta \in (\theta_3, \theta'_3)$ .
4. After Step 3 we have an interval  $(\theta_3, \theta'_3) \subseteq (0, \pi/3)$  for which there is a  $\bar{\theta}$  with the desired properties and such that  $\ell_1^*(\bar{\theta}), \ell_2^*(\bar{\theta}),$  and  $\ell_3^*(\bar{\theta})$  each lie on a fixed edge of  $L_{\text{med}}$ . Let  $v_1, v_2,$  and  $v_3$  denote the voters whose dual lines contain these three edges. We know that for any  $\theta \in (\theta_3, \theta'_3)$ , the line through  $v_1$  with orientation  $\theta$  is a balanced line. Similarly, for any  $\theta \in (\theta_3, \theta'_3)$  the line through  $v_2$  with orientation  $\theta + \pi/3$  is a balanced line, and the line through  $v_3$  with orientation  $\theta + 2\pi/3$  is a balanced line. Finding a  $\bar{\theta} \in (\theta_3, \theta'_3)$  with the desired properties thus only requires finding a  $\bar{\theta} \in (\theta_3, \theta'_3)$  for which these three lines are concurrent. Such a  $\bar{\theta}$  is guaranteed to exist by construction, and finding it is a constant-time operation.



■ **Figure 5** The edge sets  $E_1, E_2,$  and  $E_3$  of  $L_{\text{med}}$ , the median level in  $\mathcal{A}(V^*)$ .

It remains to explain how to perform Steps 1–3. Below we describe this for Step 1; the other steps can be implemented in a similar way.

To implement Step 1 we perform a binary search over the  $x$ -coordinates of the vertices of  $\mathcal{A}(V^*)$  in the slab  $(-\infty, -1/\sqrt{3}) \times (-\infty, \infty)$ , as follows. In a generic step of this binary search we have an interval  $(\theta_{\min}, \theta_{\max})$  such that  $p_{23}(\theta)$  lies to a different side of the directed line  $\ell_1(\theta)$  for  $\theta = \theta_{\min}$  than for  $\theta = \theta_{\max}$ . (Recall that this implies that there is a  $\bar{\theta} \in (\theta_{\min}, \theta_{\max})$  with the desired property.) This interval corresponds to the slab  $(\text{slope}(\theta_{\min}), \text{slope}(\theta_{\max})) \times (-\infty, \infty)$  in the dual plane. Let  $X$  be the set of  $x$ -coordinates of the vertices of  $\mathcal{A}(V^*)$  inside this slab. We can find the median  $x_{\text{med}}$  of  $X$  in  $O(n \log n)$  time using the algorithm by Cole *et al.* [8]. We then compute the three balanced lines  $\ell_i(\theta_{\text{med}})$ , where  $\theta_{\text{med}}$  is such that  $\text{slope}(\theta_{\text{med}}) = x_{\text{med}}$ . If these three lines are concurrent we are immediately done, and we can stop. Otherwise we determine where  $p_{23}(\theta_{\text{med}})$  lies relative to  $\ell_1(\theta_{\text{med}})$ , and based on that decide whether to recurse on  $(\theta_{\min}, \theta_{\text{med}})$  or on  $(\theta_{\text{med}}, \theta_{\max})$ . We continue until the slab  $(\text{slope}(\theta_{\min}), \text{slope}(\theta_{\max})) \times (-\infty, \infty)$  contains no more vertices of  $\mathcal{A}(V^*)$ . We then finish Step 1 by setting  $(\theta_1, \theta'_1) := (\theta_{\min}, \theta_{\max})$ .

Each iteration of the binary search takes  $O(n \log n)$  time, so Step 1 takes  $O(n \log^2 n)$  time. Steps 2 and 3 can be done in a similar fashion, so we can find a concurrent triple of balanced lines as in Lemma 2.4 in  $O(n \log^2 n)$  time. Next we show that the common intersection of these three lines is a  $(\sqrt{3}/2)$ -plurality point, thus proving the following lemma.

► **Lemma 2.5.** *For any finite multi-set  $V$  of voters in  $\mathbb{R}^2$  there exists a point  $p \in \mathbb{R}^2$  such that  $\beta(p, V) \geq \sqrt{3}/2$ . Moreover, such a point  $p$  can be computed in  $O(n \log^2 n)$  time.*

**Proof.** Let  $p$  be the intersection point of three concurrent balanced lines as described in Lemma 2.4—as described above we can compute these lines in  $O(n \log^2 n)$  time—and let  $q$  be a competitor. The three lines partition the plane into six equal-sized sectors, which we number  $S_1$  through  $S_6$  in a clockwise fashion, so that  $q$  lies in the closure of  $S_1$ . Let  $H$  be the closure of  $S_3 \cup S_4 \cup S_5$ . It is a closed halfspace bounded by a balanced line, so it contains at least half the voters.

Using an analysis similar to that in the proof of Lemma 2.3, we can show that  $p$  does not lose any voter  $v \in H$ . Indeed, using the Law of Sines we obtain

$$\frac{\sqrt{3}}{2} \cdot |pv| = \frac{\sqrt{3}}{2} \cdot \frac{\sin \angle pqv}{\sin \angle qpv} \cdot |qv| \leq |qv|, \quad \text{since } \angle qpv \geq \pi/3,$$

which shows that  $p$  is a  $\beta$ -plurality point for any  $\beta < \sqrt{3}/2$ . Hence,  $\beta(p, V) \geq \sqrt{3}/2$ . ◀

The following theorem summarizes the results of this section.

► **Theorem 2.6.**

- (i) *We have  $\beta_2^* = \sqrt{3}/2$ . Moreover, for any multiset  $V$  of  $n$  voters in  $\mathbb{R}^2$  we can compute a point  $p$  such that  $\beta(p, V) = \sqrt{3}/2$  in  $O(n \log^2 n)$  time.*
- (ii) *For  $d \geq 2$ , we have  $1/\sqrt{d} \leq \beta_d^* \leq \sqrt{3}/2$ . Moreover, for any multiset  $V$  of  $n$  voters in  $\mathbb{R}^d$  we can compute a point  $p$  such that  $\beta(p, V) = 1/\sqrt{d}$  in  $O(n)$  time.*

### 3 Finding a point that maximizes $\beta(p, V)$

We know from Theorem 2.6 that, for any multiset  $V$  of  $n$  voters in  $\mathbb{R}^d$ , we can compute a point  $p$  with  $\beta(p, V) \geq 1/\sqrt{d}$  (even with  $\beta(p, V) \geq \sqrt{3}/2$ , in the plane). However, a given voter multiset  $V$  may admit a  $\beta$ -plurality point for larger values of  $\beta$ —possibly even for  $\beta = 1$ . In this section we study the problem of computing a point  $p$  that maximizes  $\beta(p, V)$ , that is, a point  $p$  with  $\beta(p, V) = \beta(V)$ .

### 3.1 An exact algorithm

Below we sketch an exact algorithm to compute  $\beta(V)$  together with a point  $p$  such that  $\beta(p, V) = \beta(V)$ . Our goal is to show that, for constant  $d$ , this can be done in polynomial time. We do not make a special effort to optimize the exponent in the running time; it may be possible to speed up the algorithm, but it seems clear that it will remain impractical, because of the asymptotic running time, and also because of algebraic issues.

Note that we can efficiently check whether a true plurality point exists (i.e.,  $\beta = 1$  can be achieved) in time  $O(n \log n)$  by an algorithm of De Berg *et al.* [4], and if so, identify this point. Therefore, hereafter  $\beta = 1$  is used as a sentinel value, and our algorithm proceeds on the assumption that  $\beta(p, V) < 1$  for any point  $p$ .

For a voter  $v \in V$ , a candidate  $p \in \mathbb{R}^d$ , and an alternative candidate  $q \in \mathbb{R}^d$ , define  $f_v(p, q) := \min(|qv|/|pv|, 1)$  when  $p \neq v$ , and define  $f_v(p, q) := 1$  otherwise. Observe that for  $f_v(p, q) < 1$  we have

- $q$  wins voter  $v$  over  $p$  if and only if  $\beta > f_v(p, q)$ ,
- $q$  and  $p$  have a tie over voter  $v$  if and only if  $\beta = f_v(p, q)$ , and
- $p$  wins voter  $v$  over  $q$  if and only if  $\beta < f_v(p, q)$ .

For  $f_v(p, q) = 1$  this is not quite true: when  $p = q = v$  we always have a tie, and when  $|pv| < |qv|$  then  $p$  wins  $v$  even when  $\beta = f_v(p, q) = 1$ . When  $p = q$  there is a tie for all voters, so the final conclusion (namely that  $|V[p \succ_{\beta} q]| \geq |V[p \prec_{\beta} q]|$ ) is still correct. The fact that we incorrectly conclude that there is a tie when  $|pv| < |qv|$  and  $\beta = f_v(p, q) = 1$  does not present a problem either, since we assume  $\beta(p, V) < 1$ . Hence, we can pretend that checking if  $\beta > f_v(p, q)$ , or  $\beta = f_v(p, q)$ , or  $\beta < f_v(p, q)$  tells us whether  $q$  wins  $v$ , or there's a tie, or  $p$  wins  $v$ , respectively.

Hereafter we identify  $f_v: \mathbb{R}^{2d} \rightarrow \mathbb{R}$  with its graph  $\{(p, q, f_v(p, q))\} \subset \mathbb{R}^{2d+1}$ , which is a  $d$ -dimensional surface. Let  $f_v^+$  be the set of points lying above this graph, and  $f_v^-$  be the set of points lying below it. Thus  $f_v^+$  is precisely the set of combinations of  $(p, q, \beta)$  where  $q$  wins  $v$  over  $p$ , while  $f_v$  is the set where  $p$  ties with  $q$ , and  $f_v^-$  is the set where  $q$  loses  $v$  to  $p$ . Consider the arrangement  $\mathcal{A} := \mathcal{A}(F)$  defined by the set of surfaces  $F := \{f_v : v \in V\}$ . Each face  $C$  in  $\mathcal{A}$  is a maximal connected set of points with the property that all points of  $C$  are contained in, lie below, or lie above, the same subset of surfaces of  $F$ . (Note that we consider faces of all dimensions, not just full-dimensional cells.) Thus for all  $(p, q, \beta) \in C$ , exactly one of the following holds:  $|V[p \succ_{\beta} q]| < |V[p \prec_{\beta} q]|$ , or  $|V[p \succ_{\beta} q]| = |V[p \prec_{\beta} q]|$ , or  $|V[p \succ_{\beta} q]| > |V[p \prec_{\beta} q]|$ . Let  $L$  be the union of all faces  $C$  of  $\mathcal{A}(F)$  such that  $|V[p \succ_{\beta} q]| < |V[p \prec_{\beta} q]|$ , that is, such that  $p$  loses against  $q$  for all  $(p, q, \beta)$  in  $C$ . We can construct  $\mathcal{A}$  and  $L$  in time  $O(n^{2d+1})$  using standard machinery, as  $\mathcal{A}$  is an arrangement of degree-4 semi-algebraic surfaces of constant description complexity [2, 3]. We are interested in the set

$$W := \{(p, \beta) : |V[p \succ_{\beta} q]| \geq |V[p \prec_{\beta} q]| \text{ for any competitor } q\} \subset \mathbb{R}^{d+1}.$$

What is the relationship between  $W$  and  $L$ ? A point  $(p, \beta)$  is in  $W$  precisely when, for every choice of  $q \in \mathbb{R}^d$ ,  $p$  wins at least as many voters as  $q$  (for the given  $\beta$ ). In other words,

$$W = \{(p, \beta) \mid \text{there is no } q \text{ such that } (p, q, \beta) \in L\}.$$

That is,  $W$  is the complement of the projection of  $L$  to the space  $\mathbb{R}^{d+1}$  representing the pairs  $(p, \beta)$ . The most straightforward way to implement the projection would involve constructing semi-algebraic formulas describing individual faces and invoking quantifier elimination on the resulting formulas [2]. Below we outline a more obviously polynomial-time alternative.

Construct the *vertical decomposition*  $\text{vd}(\mathcal{A})$  of  $\mathcal{A}$ , which is a refinement of  $\mathcal{A}$  into pieces (“subfaces”  $\tau$ ), each bounded by at most  $2(2d + 1)$  surfaces of constant degree and therefore of constant complexity; see Appendix A. A vertical decomposition is specified by ordering the coordinates—we put the coordinates corresponding to  $q$  *last*. Since  $\text{vd}(\mathcal{A})$  is a refinement of  $\mathcal{A}$ , the set  $L$  is the union of subfaces  $\tau$  of  $\text{vd}(\mathcal{A})$  fully contained in  $L$ . Since  $\mathcal{A}$  is an arrangement of  $n$  well-behaved surfaces in  $2d + 1 \geq 5$  dimensions, the complexity of  $\text{vd}(\mathcal{A})$  is  $O(n^{2(2d+1)-4+\varepsilon}) = O(n^{4d-2+\varepsilon})$ , for any  $\varepsilon > 0$  [21]. In particular,  $L$  comprises  $\ell := O(n^{4d-2+\varepsilon})$  subfaces.

Since each  $\tau \subset L$  is a subspace of the vertical decomposition  $\text{vd}(\mathcal{A})$  in which the last  $d$  coordinates correspond to  $q$ , the projection  $\tau'$  of  $\tau$  to  $\mathbb{R}^{d+1}$  is easily obtained (see Appendix A) in constant time; indeed it can be obtained by discarding the constraints on these last  $d$  coordinates from the description of  $\tau$ . Thus, in time  $O(\ell)$  we can construct the family of all the projections of the  $\ell$  subfaces of  $L$ , each a constant-complexity semi-algebraic object in  $\mathbb{R}^{d+1}$ . We now construct the arrangement  $\mathcal{A}'$  of the resulting collection and its vertical decomposition  $\text{vd}(\mathcal{A}')$ . The complexity of  $\text{vd}(\mathcal{A}')$  is either  $O(\ell^{d+1+\varepsilon})$  or  $O(\ell^{2(d+1)-4+\varepsilon}) = O(\ell^{2d-2+\varepsilon})$ , depending on whether  $d + 1 \leq 4$  or not, respectively [21]. Each subspace in  $\text{vd}(\mathcal{A}')$  is either fully contained in the projection of  $L$  or fully disjoint from it. Collecting all of the latter subspaces, we obtain a representation of  $W$  as a union of at most  $O(\ell^{O(d)}) = O(n^{O(d^2)})$  constant-complexity semi-algebraic objects.

Now if  $(p, \beta) \in W$  is the point with the highest value of  $\beta$ , then  $\beta(V) = \beta(p, V) = \beta$ . It can be found by enumerating all the subspaces of  $\text{vd}(\mathcal{A}')$  contained in the closure of  $W$ —we take the closure because  $V(p, \beta)$  is defined as a supremum—and identifying their topmost point or points. Since each face has constant complexity, this can be done in  $O(1)$  time per subspace.<sup>3</sup> This completes our description of an  $O(n^{O(d^2)})$ -time algorithm to compute the best  $\beta$  that can be achieved for a given set of voters  $V$ , and the candidate  $p$  (or the set of candidates) that achieve this value.

### 3.2 An approximation algorithm

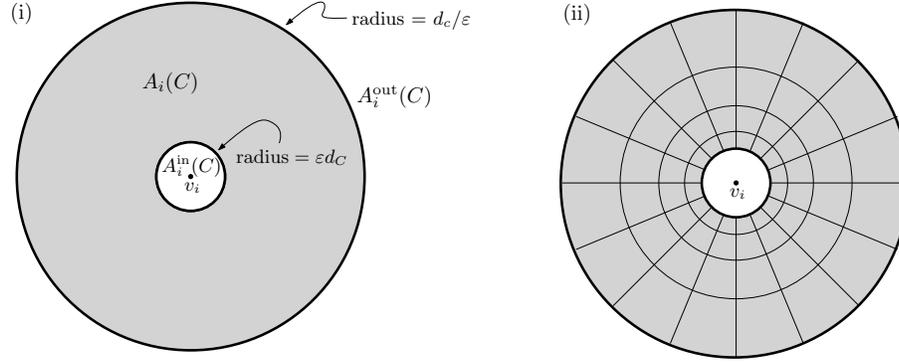
Since computing  $\beta(V)$  exactly appears expensive, we now turn our attention to approximation algorithms. In particular, given a voter set  $V$  in  $\mathbb{R}^d$  and an  $\varepsilon \in (0, 1/2]$ , we wish to compute a point  $\bar{p}$  such that  $\beta(\bar{p}, V) \geq (1 - \varepsilon) \cdot \beta(V)$ .

Our approximation algorithm works in two steps. In the first step, we compute a set  $P$  of  $O(n/\varepsilon^{2d-1} \log(1/\varepsilon))$  candidates.  $P$  may not contain the true optimal point  $p$ , but we will ensure that  $P$  contains a point  $\bar{p}$  such that  $\beta(\bar{p}, V) \geq (1 - \varepsilon/2) \cdot \beta(V)$ . In the second step, we approximate  $\beta(p', V)$  for each  $p' \in P$ , to find an approximately best candidate.

**Constructing the candidate set  $P$ .** To construct the candidate set  $P$ , we will generate, for each voter  $v_i \in V$ , a set  $P_i$  of  $O(1/\varepsilon^{2d-1} \log(1/\varepsilon))$  candidate points. Our final set  $P$  of candidates will be the union of the sets  $P_1, \dots, P_n$ . Next we describe how to construct  $P_i$ .

Partition  $\mathbb{R}^d$  into a set  $\mathcal{C}$  of  $O(1/\varepsilon^{d-1})$  simplicial cones with apex at  $v_i$  and opening angle  $\varepsilon/(2\sqrt{d})$ , so that for every pair of points  $u$  and  $u'$  in the same cone we have  $\angle uv_i u' \leq \varepsilon/(2\sqrt{d})$ . We assume for simplicity (and can easily guarantee) that no voter in  $V$  lies on the boundary of any of the cones, except for  $v_i$  itself and any voters coinciding with  $v_i$ . Let  $\mathcal{C}(v_i)$  denote

<sup>3</sup> Once again, the projection to the  $\beta$  coordinate is particularly easy to obtain if, when constructing  $\text{vd}(\mathcal{A}')$ , we set the coordinate corresponding to  $\beta$  *first*.



■ **Figure 6** (i) The closed spherical shell  $A_i(C)$  defined by the two balls of radii  $\varepsilon \cdot d_C$  and  $d_C/\varepsilon$  around  $v_i$ . (ii) The exponential grid  $G_i(C)$ . The grid is defined by a collection of spheres centered at  $v_i$ , plus extreme rays of the cones with apex at  $v_i$ . The spheres have radii  $(1 + \varepsilon/4)^i \cdot \varepsilon \cdot d_C$  for  $0 \leq i \leq \log_{(1+\varepsilon/4)}(1/\varepsilon^2) = O((1/\varepsilon) \log(1/\varepsilon))$ , and the interior angle of a cone is  $\varepsilon/2\sqrt{d}$ .

the set of all cones in  $\mathcal{C}$  whose interior contains at least one voter. For each cone  $C \in \mathcal{C}(v_i)$  we generate a candidate set  $G_i(C)$  as explained next, and then we set  $P_i := \bigcup_{C \in \mathcal{C}(v_i)} G_i(C) \cup \{v_i\}$ .

Let  $d_C$  be the distance from  $v_i$  to the nearest other voter (not coinciding with  $v_i$ ) in  $C$ . Let  $A_i(C)$  be the closed spherical shell defined by the two spheres of radii  $\varepsilon \cdot d_C$  and  $d_C/\varepsilon$  around  $v_i$ , as shown in Fig. 6(i). The open ball of radius  $\varepsilon \cdot d_C$  is denoted by  $A_i^{in}(C)$ , and the complement of the closed ball of radius  $d_C/\varepsilon$  is denoted by  $A_i^{out}(C)$ . Let  $G_i(C)$  be the vertices in an exponential grid defined by a collection of spheres centered at  $v_i$ , and the extreme rays of the cones in  $\mathcal{C}$ ; see Fig. 6(ii). The spheres have radii  $(1 + \varepsilon/4)^i \cdot \varepsilon \cdot d_C$ , for  $0 \leq i \leq \log_{(1+\varepsilon/4)}(1/\varepsilon^2) = O((1/\varepsilon) \log(1/\varepsilon))$ . Observe that  $G_i(C)$  contains not only points in  $C$ , but in the entire spherical shell  $A_i(C)$ . The set  $G_i(C)$  consists of  $O(1/\varepsilon^d \log(1/\varepsilon))$  points, and it has the following property:

Let  $p$  be any point in the spherical shell  $A_i(C)$ , and let  $p'$  be a corner of the grid cell containing  $p$  and nearest to  $p$ . Then  $|p'p| \leq \varepsilon \cdot |pv_i|$ . (\*)

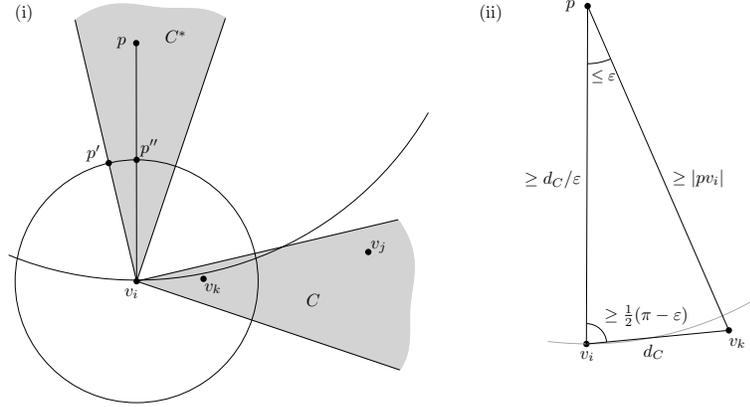
To prove the property, let  $q$  be the point on  $pv_i$  such that  $|qv_i| = |p'v_i|$ . From the construction of the exponential grid we have  $|pq| \leq \frac{\varepsilon}{4} \cdot |pv_i|$ . Since  $p'$  and  $q$  lie in the same cone  $\angle p'v_iq \leq \frac{\varepsilon}{2\sqrt{d}}$  and, consequently,  $|p'q| \leq \frac{\varepsilon}{2} \cdot |qv_i| \leq (1 + \frac{\varepsilon}{4}) \cdot \frac{\varepsilon}{2} \cdot |pv_i|$ . The property is now immediate since  $|pp'| \leq |pq| + |qp'| < \varepsilon \cdot |pv_i|$ .

As mentioned above,  $P_i := \bigcup_{C \in \mathcal{C}(v_i)} G_i(C) \cup \{v_i\}$ , and the final candidate set  $P$  is defined as  $P := \bigcup_{v_i \in V} P_i$ . Computing the sets  $P_i$  is easy: for each of the  $O(1/\varepsilon^{d-1})$  cones  $C \in \mathcal{C}(v_i)$ , determine the nearest neighbor of  $v_i$  in  $C$  in  $O(n)$  time by brute force, and then generate  $G_i(C)$  in  $O((1/\varepsilon^{(d-1)}) \log(1/\varepsilon))$  time. (It is not hard to speed up the nearest-neighbor computation using appropriate data structures, but this will not improve the final running time in Theorem 3.4.) We obtain the following lemma.

► **Lemma 3.1.** *The candidate set  $P$  has size  $O(n/\varepsilon^{2d-1} \log(1/\varepsilon))$  and can be constructed in  $O(n^2/\varepsilon^{d-1} + n/\varepsilon^{2d-1} \log(1/\varepsilon))$  time.*

The next lemma is crucial to show that  $P$  is a good candidate set.

► **Lemma 3.2.** *For any point  $p \in \mathbb{R}^d$ , there exists a point  $p' \in P$  with the following property: for any voter  $v_j \in V$ , we have that  $|p'v_j| \leq (1 + 2\varepsilon) \cdot |pv_j|$ .*



■ **Figure 7** Illustration for Case III.

**Proof.** Let  $v_i$  be a voter nearest to  $p$ . We will argue that the set  $P_i$  contains a point  $p'$  with the desired property. We distinguish three cases.

*Case I:* There is a cone  $C \in \mathcal{C}(v_i)$  such that  $p$  lies in the spherical shell  $A_i(C)$ . In this case we pick  $p'$  to be a point of  $G_i(C)$  nearest to  $p$ , that is,  $p'$  is a corner nearest to  $p$  of the grid cell containing  $p$ . By property (\*) we have

$$|p'v_j| \leq |p'p| + |pv_j| \leq \varepsilon \cdot |pv_i| + |pv_j| \leq (1 + \varepsilon) \cdot |pv_j|,$$

where the last inequality follows from the fact that  $v_i$  is a voter nearest to  $p$ .

*Case II:* Point  $p$  lies in  $A_i^{\text{in}}(C)$  for all  $C \in \mathcal{C}(v_i)$ . In this case we pick  $p' := v_i$ . Clearly  $|p'v_j| = 0 \leq (1 + \varepsilon) \cdot |pv_j|$  for  $j = i$ . For  $j \neq i$ , we argue as follows. Let  $C \in \mathcal{C}(v_i)$  be the cone containing  $v_j$ . Since we are in Case II we know that  $p \in A_i^{\text{in}}(C)$ , and so

$$|p'v_j| \leq |p'p| + |pv_j| \leq \varepsilon d_C + |pv_j| \leq \varepsilon |p'v_j| + |pv_j|. \quad (1)$$

Moreover, we have

$$|pv_j| \geq |p'v_j| - |pp'| \geq |p'v_j| - \varepsilon d_C \geq |p'v_j|/2, \quad (2)$$

where the last step uses that  $\varepsilon \leq 1/2$  and  $d_C \leq |p'v_j|$ . Combining (1) and (2), we obtain  $|p'v_j| \leq (1 + 2\varepsilon) \cdot |pv_j|$ .

*Case III:* Cases I and II do not apply. In this case there is at least one cone  $C$  such that  $p \in A_i^{\text{out}}(C)$ . Of all such cones, let  $C^*$  be the one whose associated distance  $d_{C^*}$  is maximized. Let  $p''$  be the point on the segment  $pv_i$  at distance  $d_C/\varepsilon$  from  $v_i$ . Without loss of generality, we will assume that  $p$  and  $v_i$  only differ in the  $x_d$  coordinate; see Fig. 7(i).

We will prove that the point  $p'$  of  $G_i(C^*)$  nearest to  $p''$  (refer to Fig. 7(i)) has the desired property. Consider a voter  $v_j$ . We distinguish three cases.

- When  $i = j$ , then we have

$$|p'v_i| \leq |p'p''| + |p''v_i| \leq (1 + \varepsilon)|p''v_i| \leq (1 + \varepsilon)|pv_i|,$$

where the second inequality follows from (\*).

- When  $v_j$  lies in a cone  $C$  such that  $p \in A_i^{\text{in}}(C)$ , then we can use the same argument as in Case II to show that  $|p'v_j| \leq (1 + 2\varepsilon) \cdot |pv_j|$ .

- In the remaining case  $v_j$  lies in a cone  $C$  such that  $p \in A_i^{\text{out}}(C)$ . Let  $v_k$  be a voter in  $C$  nearest to  $v_i$ . Since  $|v_i v_k| = d_C$ ,  $|pv_i| \geq d_C/\varepsilon$ , and  $|pv_k| \geq |pv_i|$ , we can deduce that  $\angle pv_i v_k \geq \pi/2 - \varepsilon/2$ , as illustrated in Fig. 7(ii). Furthermore, since  $v_k$  and  $v_j$  belong to the same cone  $C$  the angle  $\angle v_k v_i v_j$  is bounded by  $\varepsilon/2\sqrt{d} \leq \varepsilon/2$  according to the construction. Putting the two angle bounds together we conclude that  $\angle pv_i v_j \geq \frac{\pi}{2} - \varepsilon$ . Now consider the triangle defined by  $p, v_i$  and  $v_j$ . From the Law of Sines we obtain

$$\frac{|v_i v_j|}{\sin \angle v_i p v_j} = \frac{|pv_j|}{\sin \angle p v_i v_j}, \quad \text{or} \quad |v_i v_j| = |pv_j| \cdot \frac{\sin \angle v_i p v_j}{\sin \angle p v_i v_j} \leq \frac{|pv_j|}{\cos \varepsilon} \leq (1 + \varepsilon) \cdot |pv_j|,$$

for  $\varepsilon < 1/2$ . Since  $p''$  lies on the line between  $p$  and  $v_i$  we have:

$$|p'' v_j| \leq \max\{|pv_j|, |v_i v_j|\} \leq (1 + \varepsilon) \cdot |pv_j|.$$

Finally we get the claimed bound by noting that  $|p' p''| \leq \varepsilon \cdot |p' v_i|$  (from (\*)),

$$|p' v_j| \leq |p' p''| + |p'' v_j| \leq \varepsilon \cdot |p' v_i| + (1 + \varepsilon) \cdot |pv_j| \leq (1 + 2\varepsilon) \cdot |pv_j|. \quad \blacktriangleleft$$

**An approximate decision algorithm.** Given a point  $p$ , a positive real value  $\varepsilon$  and the voter multiset  $V$ , we say that an algorithm ALG is an  $\varepsilon$ -approximate decision algorithm if

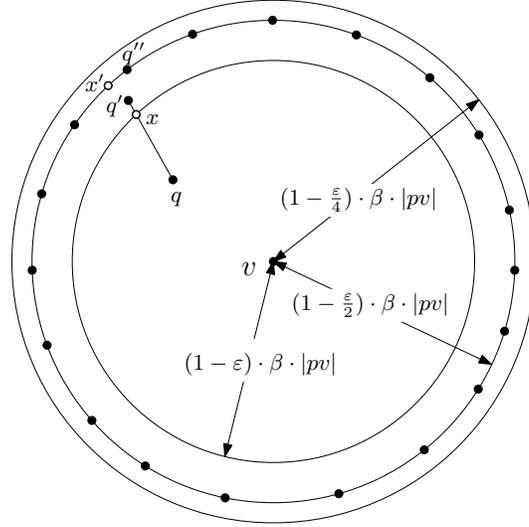
- ALG answers YES if  $p$  is a  $\beta$ -plurality point, and
- ALG answers NO if  $p$  is not a  $(1 - \varepsilon)\beta$ -plurality point.

In the remaining cases, where  $(1 - \varepsilon)\beta < \beta(p, V) < \beta$ , ALG may answer YES or NO.

Next we propose an  $\varepsilon$ -approximate decision algorithm ALG. The algorithm will use the so-called Balanced Box-Decomposition (BBD) tree introduced by Arya and Mount [1]. BBD trees are hierarchical space-decomposition trees such that each node  $\mu$  represents a region in  $\mathbb{R}^d$ , denoted by  $\text{region}(\mu)$ , which is a  $d$ -dimensional axis-aligned box or the difference of two such boxes. A BBD tree for a set  $P$  of  $n$  points in  $\mathbb{R}^d$  can be built in  $O(n \log n)$  time using  $O(n)$  space. It supports  $(1 + \varepsilon)$ -approximate range counting queries with convex query ranges in  $O(\log n + \varepsilon^{1-d})$  time [1]. In our algorithm all query ranges will be balls, hence a  $(1 + \varepsilon)$ -approximate range-counting query for a  $d$ -dimensional ball  $s(v, r)$  with center at  $v$  and radius  $r$  returns an integer  $I$  such that  $|P \cap s(v, r)| \leq I \leq |P \cap s(v, (1 + \varepsilon)r)|$ .

Our  $\varepsilon$ -approximate decision algorithm ALG works as follows.

1. Construct a set  $Q$  of  $O(n/\varepsilon^{d-1})$  potential candidates competing against  $p$ , as follows. Let  $Q(v)$  be a set of  $O(1/\varepsilon^{d-1})$  points distributed uniformly on the boundary of the ball  $s(v, (1 - \varepsilon/2) \cdot \beta \cdot |pv|)$ , such that the distance between any point on the boundary and its nearest neighbor in  $Q(v)$  is at most  $\frac{\varepsilon}{4\sqrt{d}} \cdot |pv| \leq \frac{\varepsilon}{4} \cdot \beta \cdot |pv|$ . In the last step we use the fact that  $\beta \geq 1/\sqrt{d}$ , according to Lemma 2.3. Set  $Q := Q(v_1) \cup \dots \cup Q(v_n)$ .
2. Build a BBD tree  $\mathcal{T}$  on  $Q$ . Add a counter  $c(\mu)$  to each node  $\mu$  in  $\mathcal{T}$ , initialized to zero.
3. For each voter  $v \in V$  perform a  $(1 + \varepsilon/4)$ -approximate range-counting query with  $s(v, (1 - \varepsilon/4) \cdot \beta \cdot |pv|)$  in  $\mathcal{T}$ . We modify the search in  $\mathcal{T}$  slightly as follows. If an internal node  $\mu \in \mathcal{T}$  is visited and expanded during the search, then for every non-expanded child  $\mu'$  of  $\mu$  with  $\text{region}(\mu')$  entirely contained in  $s(v, (1 + \varepsilon/4)(1 - \varepsilon/4) \cdot \beta \cdot |pv|) \subset s(v, \beta \cdot |pv|)$  we increment the counter  $c(\mu')$ . Similarly, if a leaf is visited then the counter is incremented if the point stored in the leaf lies within  $s(v, (1 - \varepsilon/4) \cdot \beta \cdot |pv|)$ .
4. For a leaf  $\mu$  in  $\mathcal{T}$ , let  $M(\mu)$  be the set of nodes in  $\mathcal{T}$  on the path from the root to  $\mu$ , and let  $C(\mu) = \sum_{\mu' \in M(\mu)} c(\mu')$ . Compute  $C(\mu)$  for all leaves  $\mu$  in  $\mathcal{T}$  by a pre-order traversal of  $\mathcal{T}$ , and set  $C := \max_{\mu} C(\mu)$ .
5. If  $C \leq n/2$ , then return YES, otherwise NO.



■ **Figure 8** Illustrating the three balls of different radius used in the correctness proof of Lemma 3.3.

To prove correctness of the algorithm we define, for a given  $\gamma > 0$ , a *fuzzy ball*  $s_\gamma(v, r)$  to be any set such that  $s(v, r) \subseteq s_\gamma(v, r) \subseteq s(v, (1 + \gamma)r)$ . Thus if  $q \in s(v, r)$  then  $q \in s_\gamma(v, r)$ , if  $q \notin s(v, (1 + \gamma)r)$  then  $q \notin s_\gamma(v, r)$ , and otherwise  $q$  may or may not be inside in  $s_\gamma(v, r)$ . We now observe that for each voter  $v_i \in V$  there is a fuzzy ball  $s_{\varepsilon/4}(v_i, (1 - \varepsilon/4) \cdot \beta \cdot |pv_i|)$  such that the value  $C(\mu)$  for a leaf  $\mu$  storing a point  $q$  is the depth of  $q$  in the arrangement, denoted by  $\mathcal{A}_{\varepsilon/4}(V, 1 - \varepsilon/4)$ , of the fuzzy balls  $s_{\varepsilon/4}(v_1, (1 - \varepsilon/4) \cdot \beta \cdot |pv_1|), \dots, s_{\varepsilon/4}(v_n, (1 - \varepsilon/4) \cdot \beta \cdot |pv_n|)$ .

► **Lemma 3.3.** *Algorithm ALG  $\varepsilon$ -approximately decides if  $p$  is a  $\beta$ -plurality point in time  $O(\frac{n}{\varepsilon^{d-1}} \log \frac{n}{\varepsilon^{d-1}})$ .*

**Proof.** We start by analyzing the running time of the algorithm. Constructing the set of points in  $Q$  can be done in time linear in  $|Q|$ , while building the BBD-tree  $\mathcal{T}$  requires  $O((n/\varepsilon^{d-1}) \log(n/\varepsilon^{d-1}))$  time [1, Lemma 1]. Next, the algorithm performs  $n$  approximate range queries, each requiring  $O(\log \frac{n}{\varepsilon^{d-1}} + \frac{1}{\varepsilon^{d-1}})$  time [1, Theorem 2]). Note that the small modification we made to the query algorithm to update the counters does not increase the asymptotic running time. Finally, the traversal of  $\mathcal{T}$  to compute  $C$  takes time linear in the size of  $\mathcal{T}$ , which is  $O(n/\varepsilon^{d-1})$ .

It remains to prove that ALG is correct.

- If  $p$  is a plurality point there can be no point  $q \in \mathbb{R}^d$  having depth greater than  $n/2$  in the arrangement of the balls  $s(v_1, \beta \cdot |pv_1|), \dots, s(v_n, \beta \cdot |pv_n|)$ . Since  $s_{\varepsilon/4}(v, (1 - \varepsilon/4) \cdot \beta \cdot |pv|) \subset s(v, \beta \cdot |pv|)$ , for all  $v$ , ALG could not have found a point with depth greater than  $n/2$ , and hence, must return YES.
- If  $p$  is not a  $(1 - \varepsilon)\beta$ -plurality point, then there exists a point  $q$  with depth greater than  $n/2$  in the arrangement  $\mathcal{A}(V, 1 - \varepsilon)$  of the balls  $s(v_1, (1 - \varepsilon) \cdot \beta \cdot |pv_1|), \dots, s(v_n, (1 - \varepsilon) \cdot \beta \cdot |pv_n|)$ . Let  $q'$  be the point in  $Q$  nearest to  $q$ . We claim that for any ball  $s(v, (1 - \varepsilon) \cdot \beta \cdot |pv|)$  that contains  $q$ , its expanded version  $s(v, (1 - \varepsilon/4) \cdot \beta \cdot |pv|)$  contains  $q'$ . Of course, if  $s(v, (1 - \varepsilon) \cdot \beta \cdot |pv|)$  contains  $q'$  then we are done. Otherwise, let  $x$  be the point where  $qq'$  intersects the boundary of  $s(v, (1 - \varepsilon) \cdot \beta \cdot |pv|)$ ; see Fig. 8. Note that  $q'$  must also be the point in  $Q$  nearest to  $x$ .

Let  $x'$  be the point on the boundary of  $s(v, (1 - \varepsilon/2) \cdot \beta \cdot |pv|)$  nearest to  $x$ , and let  $q''$  be a point in  $Q$  on the boundary of  $s(v, (1 - \varepsilon/2) \cdot \beta \cdot |pv|)$ . By construction, we have

$$|xx'| = \frac{\varepsilon}{4} \cdot \beta \cdot |pv| \quad \text{and} \quad |x'q''| \leq \frac{\varepsilon}{4} \cdot \beta \cdot |pv|$$

and, by the triangle inequality, we obtain

$$|xq'| \leq |xq''| \leq |xx'| + |x'q''| \leq \frac{\varepsilon}{2} \cdot \beta \cdot |pv|.$$

This implies that  $s(v, (1 - \varepsilon/4) \cdot \beta \cdot |pv|) \subseteq s_{\varepsilon/4}(v, (1 - \varepsilon/4) \cdot \beta \cdot |pv|)$  must contain  $q'$ . Consequently, if  $q$  has depth at least  $n/2$  in  $\mathcal{A}(V, 1 - \varepsilon)$  then  $q'$  has depth at least  $n/2$  in the arrangement  $\mathcal{A}_{\varepsilon/4}(V, (1 - \varepsilon/4))$ , and hence, the algorithm will return NO.  $\blacktriangleleft$

**The algorithm.** Now we have the tools required to approximate  $\beta(V)$ . First, generate the set  $P$  of  $O(\frac{n}{\varepsilon^{2d-1}} \log \frac{1}{\varepsilon})$  candidate points. For each candidate point  $p \in P$ , perform a binary search for an approximate  $\beta^*(p)$  in the interval  $[1/\sqrt{d}, 1]$ , until the remaining search interval has length at most  $\varepsilon/2 \cdot 1/\sqrt{d}$ . For each  $p$  and  $\beta^*$ ,  $(\varepsilon/2)$ -approximately decide if  $p$  is a  $\beta^*$ -plurality point in  $V$ . Return the largest  $\beta^*$  and the corresponding point  $p$  on which the algorithm says YES.

► **Theorem 3.4.** *Given a multiset  $V$  of voters in  $\mathbb{R}^d$ , a  $((1 - \varepsilon) \cdot \beta(V))$ -plurality point can be computed in  $O(\frac{n^2}{\varepsilon^{3d-2}} \cdot \log \frac{n}{\varepsilon^{d-1}} \cdot \log^2 \frac{1}{\varepsilon})$ .*

## 4 Concluding Remarks

We proved that any finite set of voters in  $\mathbb{R}^d$  admits a  $\beta$ -plurality point for  $\beta = 1/\sqrt{d}$  and that some sets require  $\beta = \sqrt{3}/2$ . For  $d = 2$  we managed to close the gap by showing that  $\beta_2^* = \sqrt{3}/2$ . One of the main open problems is to close the gap for  $d > 2$ . We also presented an approximation algorithm that finds, for a given  $V$ , a  $(1 - \varepsilon) \cdot \beta(V)$ -plurality point. The algorithm runs in  $O^*(n^2/\varepsilon^{3d-2})$  time. Another open problem is whether a subquadratic approximation algorithm exists, and to prove lower bounds on the time to compute  $\beta(V)$  or  $\beta(p, V)$  exactly. Finally, it will be interesting to study  $\beta$ -plurality points in other metrics, for instance in the personalized  $L_1$ -metric [4] for  $d > 2$ .

## Acknowledgements

The authors would like to thank Sampson Wong for improving an earlier version of Lemma 2.5.

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## Appendix

### A A primer on vertical decompositions

We follow the notation and terminology of [6, 25]. A *vertical decomposition* is, roughly, any partition of space into finitely many so-called cylindrical cells (see below for a definition); it need not be a topological complex. A *vertical decomposition of an arrangement* is a refinement of an arrangement into cylindrical cells,<sup>4</sup> where *refinement* means that each cylindrical cell is a subset of a face in the arrangement. We define cylindrical cells recursively. To simplify the notation, any inequality limit in our definitions can be omitted, i.e., replaced by a  $\pm\infty$ , as appropriate. For example, when we talk about an open interval  $(a, b)$ , i.e., the set of numbers  $x$  with  $a < x < b$ , we include the possibilities of the unbounded intervals  $(-\infty, b)$ ,  $(a, +\infty)$ , and  $(-\infty, +\infty)$ .

A *one-dimensional* cylindrical cell is either a singleton or an open interval  $(a, b)$ . So a one-dimensional vertical decomposition is a decomposition of  $\mathbb{R}$  into a finite number of singletons and intervals.

We now define a cylindrical cell  $\tau$  in  $\mathbb{R}^2$ . Its projection  $\tau'$  to the  $x_1$ -axis is a cylindrical cell in  $\mathbb{R}$ . The cell  $\tau$  must have one of the following two forms:

- $\{(x_1, f_2(x_1)) \mid x_1 \in \tau'\}$ , where  $f_2 : \tau' \rightarrow \mathbb{R}$  is a continuous total function, or
- $\{(x_1, x_2) \mid x_1 \in \tau', f_2(x_1) < x_2 < g_2(x_1)\}$ , where  $f_2, g_2 : \tau' \rightarrow \mathbb{R}$  are two continuous total functions, with the property that  $f_2(x_1) < g_2(x_1)$  for all  $x_1 \in \tau'$ .

If  $\tau'$  is a singleton, the former defines a vertex and the latter an (open) vertical segment. If  $\tau'$  is an interval, the former defines an open monotone arc (a portion of the graph of the function  $f_2$ ) and the latter an open *pseudo-trapezoid* delimited by two (possibly degenerate) vertical segments on left and right and by the two disjoint function graphs below and above. (Recall that any of the limits may be omitted. For example,  $\mathbb{R}^2$  is a legal cell in a trivial two-dimensional vertical decomposition consisting only of itself, where all the limits have been “replaced by infinities.”)

A cylindrical cell  $\tau \subset \mathbb{R}^d$  is defined recursively. Its projection  $\tau'$  is a cylindrical cell in  $\mathbb{R}^{d-1}$ . Moreover,  $\tau$  must have one of the following forms:

- $\{(x', f_d(x_1, \dots, x_{d-1})) \mid x' \in \tau'\}$ , where  $f_d : \tau' \rightarrow \mathbb{R}$  is a continuous total function, or
- $\{(x', x_d) \mid x' \in \tau', f_d(x') < x_d < g_d(x')\}$ , where  $f_d, g_d : \tau' \rightarrow \mathbb{R}$  are two continuous total functions, with the property that  $f_d(x') < g_d(x')$  for all  $x' \in \tau'$ .

A cylindrical cell is fully specified by giving its dimension and the sequence of functions  $f_i$  or pairs of functions  $f_i, g_i$ , as appropriate. In particular, the projection of the cell in a  $k$ -dimensional decomposition to its first  $k' < k$  coordinates can be obtained by retaining the inequalities in the first  $k'$  coordinates and discarding the remaining ones.

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<sup>4</sup> The specific decomposition depends on the algorithm used to construct it and on the ordering of the coordinates. In the computational- and combinatorial-geometry literature, one often speaks of “the vertical decomposition of the arrangement” in the sense of “the vertical decomposition obtained by applying the algorithm, say, of Chazelle *et al.* [6] or of Koltun [21], to the given arrangement.”