

Let  $\mathbf{v} = (v_1, v_2, \dots, v_n) > 0$  be an  $n$ -dimensional vector of values,  $\mathbf{e} = (1, 1, \dots, 1)$  the  $n$  dimensional vector of all ones,  $\mathbf{a} = \mathbf{v}/(\mathbf{e}^T \mathbf{v})$  the corresponding target vector, and  $\mathbf{x}^0$  a starting vector with the property that  $\mathbf{e}^T \mathbf{x}^0 = 1$ . We seek a procedure that generates a sequence of vectors  $\mathbf{x}^0, \mathbf{x}^1, \dots$  that converges to  $\mathbf{a}$ . Notice that a vector corresponds to a point in the  $n$ th dimensional Euclidean space, and so we define the Euclidean distance  $d(\mathbf{v}^1, \mathbf{v}^2)$  between two vectors  $\mathbf{v}_1$  and  $\mathbf{v}_2$  as the Euclidean distance between the two corresponding points. In particular, we will be interested in the rate of convergence to zero of  $d(\mathbf{x}^t, \mathbf{a})$ . The  $i$ th component of  $\mathbf{x}^t$  will be denoted as  $x_i^t$ . At each stage, we are allowed to choose two components  $i$  and  $j > i$  and redistribute their value, that is,  $\mathbf{x}^{t+1} = (x_1^t, \dots, x_i^{t+1}, \dots, x_j^{t+1}, \dots, x_n^t)$ , where  $x_i^{t+1} + x_j^{t+1} = x_i^t + x_j^t \stackrel{\text{def}}{=} s_t$ . We propose the algorithm that assigns

$$x_i^{t+1} = s_t \frac{v_i}{v_i + v_j} . \quad (1)$$

**Lemma 1** *When  $n = 3$ , algorithm (1) yields*

$$E[d(\mathbf{x}^{t+1}, \mathbf{a}) | \mathbf{x}^t, \mathbf{x}^{t-1}, \dots, \mathbf{x}^0] \leq \frac{4}{3\sqrt{3}} E[d(\mathbf{x}^t, \mathbf{a}) | \mathbf{x}^{t-1}, \dots, \mathbf{x}^0] ,$$

where the expectations is taken over random choices of  $i$  and  $j$ .

**Proof.** The proof is a geometric construction of  $\mathbf{x}^{t+1}$  on the basis of  $\mathbf{x}^t$ , which yields an upper bound on  $d(\mathbf{x}^t, \mathbf{a})$  that does not depend on previous  $\mathbf{x}$ 's. First, we define the set of feasible vector and show that it is a triangle. Moreover, we will show how to construct  $\mathbf{x}^{t+1}$  from  $\mathbf{x}^t$  through an elementary geometric projection. A trigonometric analysis of such projection will yield the desired bound.

Define the set of feasible points as the triangle  $x_1 + x_2 + x_3 = 1, x_1, x_2, x_3 \geq 0$ . Observe that such triangle is equilateral and so its angles are  $\pi/3$ . Suppose first that  $i = 1, j = 2$ , and so  $x_3$  is left unchanged in the current step. Hence, the resulting vector will be a point in the segment defined by  $x_1 + x_2 + x_3 = 1, x_1, x_2, x_3 \geq 0$ , and  $x_3 = x_3^t$ . Notice that the line defined by  $x_1 + x_2 + x_3 = 1$  and  $x_3 = x_3^t$  is parallel to the side of triangle with  $x_3 = 0$ . In general, a step moves the current vector  $\mathbf{x}^t$  to a new point  $\mathbf{x}^{t+1}$  along a line which is parallel to a triangle side. Hence, there are only three lines along which movement can occur. Let  $\alpha$  be the angle between any two of those lines, and observe that  $\alpha = \pi/3$  by an elementary property of parallel lines which is depicted in figure 1. Let  $u, w$ , and  $z$  be the three lines parallel to triangle sides passing through  $\mathbf{x}^t$ , and notice that these three lines partition the plane into size equal angles. We next give simple geometric construction to determine  $\mathbf{x}^{t+1}$  assuming that the coordinate  $x_3$  is left unchanged. Consider first the triangle ACD in figure 2 and notice that such triangle is rectangle and its base is  $s_t$ , which is equal to its height. Analogously, EGH is rectangle and both its base and height are  $a_2 + a_3$ . Hence, ACD and EGH are similar. Let us turn now to  $AB\mathbf{x}^{t+1}$  and notice that it is rectangle, its base is  $x_1^{t+1}$ , and its height is  $x_2^{t+1}$ . Hence, height and base are in the ratio  $v_1/v_2$ . However, the triangle  $EG\mathbf{a}$  is also rectangle and its base and height are in the same ratio  $v_1/v_2$ . Hence,  $AB\mathbf{x}^{t+1}$  and  $EG\mathbf{a}$  are similar. Therefore, we can use the following construction to determine  $\mathbf{x}^{t+1}$ . Suppose that  $\mathbf{x}^{t+1}$  is chosen along line  $w$ , as in figure 1. Take a line  $l$  that passes through

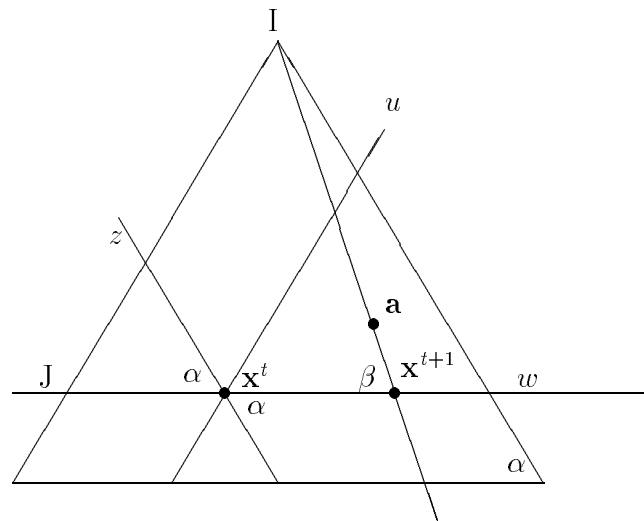


Figure 1: Geometric construction of  $\mathbf{x}^t$ .

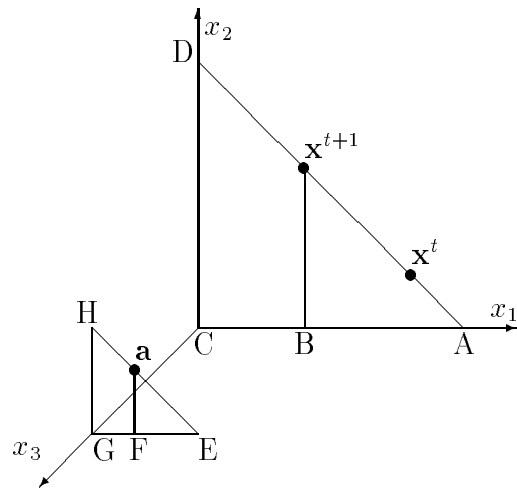


Figure 2: Similarity between triangles.

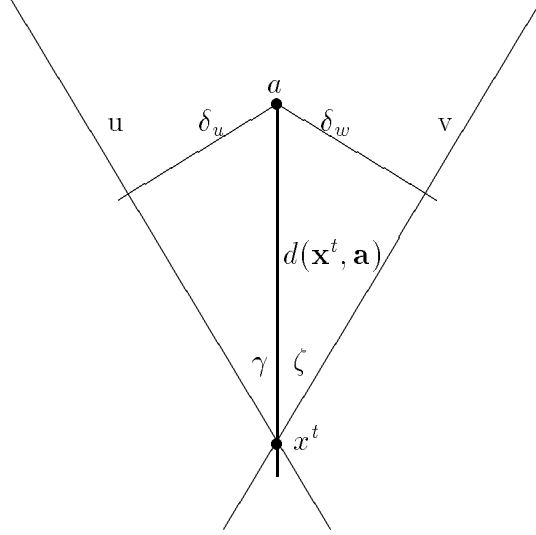


Figure 3: Relation between distances.

the vertex  $I$  opposite to  $w$  and through  $\mathbf{a}$ . Then,  $\mathbf{x}^{t+1}$  is located at the intersection of  $l$  and  $w$ .

We next turn to show a bound on the angle  $\beta$  formed by  $l$  and  $w$ , and specifically that  $\pi/3 < \beta < 2\pi/3$ . Indeed, take the line passing through  $\mathbf{x}^{t+1}$  that is parallel to  $z$ , and observe that  $\beta \geq \alpha = \pi/3$ . As for the second part of the statement, notice that the angle  $J$  is  $\pi/3$ , and so by the sum of the angles in the triangle  $IJ\mathbf{x}^{t+1}$ ,  $\beta < 2\pi/3$ . Assume without loss of generality that  $\mathbf{a}$  is in the partition determined by  $u$  and  $w$ , and denote by  $\delta_w$  ( $\delta_u, \delta_z$ ) the distance between  $\mathbf{a}$  and  $w$  ( $u, z$ ). We will show that the  $\delta$ 's are not much smaller than  $d(\mathbf{x}^{t+1}, \mathbf{a})$ . Specifically, if  $\mathbf{x}$  is chosen along  $w$ , then  $d(\mathbf{x}^{t+1}, \mathbf{a}) = \delta_w \sin \beta \leq 2\delta_w/\sqrt{3}$  because  $\pi/3 \leq \beta < 2\pi/3$ . Analogous results hold also for the other lines. We now turn to estimate an upper bound to the expected value of  $\delta$  for a random choice of  $i$  and  $j$ . First, notice that  $\delta_z \leq d(\mathbf{x}^t, \mathbf{a})$  by definition of  $\delta_z$ . Moreover, the line  $\mathbf{x}^t\mathbf{a}$  partitions a  $\pi/3$  angle into angles  $\gamma$  and  $\zeta$  with  $\gamma + \zeta = \pi/3$  as in figure 3. Hence,  $\delta_u = d(\mathbf{x}^t, \mathbf{a}) \sin \gamma$ ,  $\delta_w = d(\mathbf{x}^t, \mathbf{a}) \sin \zeta = d(\mathbf{x}^t, \mathbf{a}) \sin(\pi/3 - \gamma)$ . Hence,  $\delta_u + \delta_w = d(\mathbf{x}^t, \mathbf{a}) \sin(\gamma + \pi/3) \leq d(\mathbf{x}^t, \mathbf{a})$ . On the whole, the expected  $\delta$  is  $(\delta_u + \delta_w + \delta_z)/3 \leq 2d(\mathbf{x}^t, \mathbf{a})$ , so that the expected  $d(\mathbf{x}^{t+1}, \mathbf{a}) \leq 4/3\sqrt{3}d(\mathbf{x}^t, \mathbf{a})$ . We notice that such bound is independent of the previous values of  $\mathbf{x}^t, \mathbf{x}^{t-1}, \dots, \mathbf{x}^0$ , and the lemma is proven.  $\square$

The repeated application of the previous lemma yields that  $\mathbf{x}^t$  converges exponentially

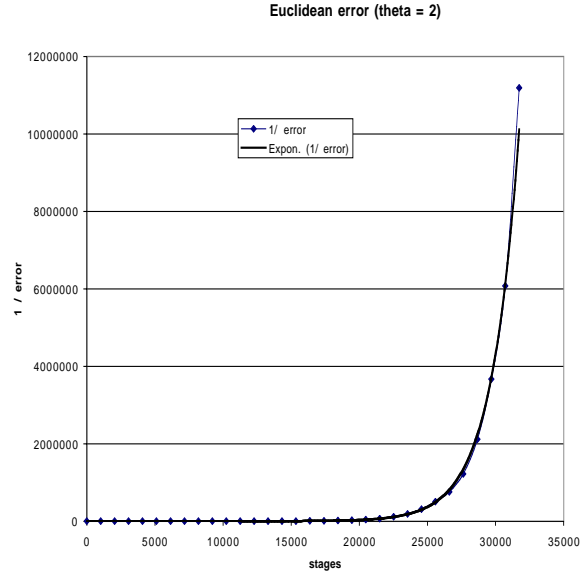


Figure 4:  $n = 1024$  for  $\theta = 2$ .

fast to  $\mathbf{a}$ .

Our experimental results suggest that the proposition generalizes when  $n > 3$ , as shown in figure 4. The convergence has is given for  $n = 1024$ , where  $v_i = i^{-\theta}$  for  $\theta = 2$ .