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# A Subquadratic Algorithm for Computing the $L_1$ -distance between Two Terrains

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## Abstract

We study the problem of computing the  $L_1$ -distance between two piecewise-linear bivariate functions  $f$  and  $g$ , defined over a common bounded domain  $\mathbb{M} \subset \mathbb{R}^2$ , that is, computing the quantity  $\|f - g\|_1 = \int_{\mathbb{M}} |f(x, y) - g(x, y)| dx dy$ . If  $f$  and  $g$  are defined by linear interpolation over triangulations  $\mathbf{T}_f$  and  $\mathbf{T}_g$ , respectively, of  $\mathbb{M}$  with a total of  $n$  triangles, we show that  $\|f - g\|_1$  can be computed in  $\tilde{O}(n^{(\omega+1)/2})$  time, where  $\Theta(n^\omega)$  is the time required to multiply two  $n \times n$  matrices and  $\tilde{O}$  notation hides polylogarithmic factors; this bound holds for the currently best known value of  $\omega$ , which is approximately 2.37. The previously best known algorithm for computing  $\|f - g\|_1$  takes  $\Theta(n^2)$  time in the worst case.

More generally, if the complexity of the overlay of  $\mathbf{T}_f$  and  $\mathbf{T}_g$  is  $\kappa$ , then the runtime of our algorithm is  $\tilde{O}(\kappa^{(\omega-1)/2} n^{(3-\omega)/2})$ .

## 1 Introduction

Combining tools from computational geometry and computer algebra has led to new results that defy intuitive lower bounds. For example, using algebraic tools, it is possible to compute the centroid of the vertices in an arrangement of  $n$  lines in the plane in  $O(n \log^2 n)$  operations, without computing the  $\Theta(n^2)$  vertices [5]. One notable algebraic tool used for this problem is an algorithm to evaluate efficiently a univariate polynomial on multiple points [11]. This tool was used to speed up, among others, the computation of Shapley values in the plane [8]; the location of points among wireless transmitters [6]; and the computation of the  $L_2$ -distance between two piecewise-linear bivariate functions  $f$  and  $g$ , see below for the definition [13].

Bivariate functions over a planar domain appear naturally in geographical information systems, under the name of *terrains*, where they are used to model actual geographic terrains, such as elevation in a hilly area, and more generally to represent two-dimensional datasets, such as precipitation, air pollution, population density, or quantities of wine production. One way of approximating such a function is to triangulate the underlying domain and define a linear function over each triangle, resulting in a piecewise-linear function, often referred to as a *triangulated irregular network*. In this paper, we design an efficient algorithm for computing the  $L_1$ -distance between two piecewise-linear function, defined below, by combining tools from computational geometry and computer algebra.

**Problem statement.** Let  $\mathbb{M}$  be a bounded polygonal region of  $\mathbb{R}^2$ , and let  $f, g : \mathbb{M} \rightarrow \mathbb{R}$  be two piecewise-linear functions defined over their common domain  $\mathbb{M}$ . Each of  $f, g$  is associated with a triangulation of  $\mathbb{M}$ , denoted by  $\mathbf{T}_f, \mathbf{T}_g$ , respectively, and  $f$  (resp.,  $g$ ) is defined to be linear over each triangle of  $\mathbf{T}_f$  (resp.,  $\mathbf{T}_g$ ). Let  $n$  be the total number of vertices in the two triangulations. For any integer  $p > 0$ , we define the  $L_p$ -distance between  $f$  and  $g$  to be

$$\|f - g\|_p = \left( \int_{\mathbb{M}} |f - g| dx dy \right)^{1/p}. \quad (1)$$

The goal in this paper is to compute  $\|f - g\|_1$ .

We will work in the Real RAM model [15], where inputs and intermediate values are real numbers and algebraic operations and comparisons can be performed in constant time, but modulus or rounding to integers operations are not allowed. We assume that the coefficients of the piecewise linear functions  $f$  and  $g$  and the coordinates of the triangles in  $\mathbf{T}_f$  and  $\mathbf{T}_g$  are real numbers.

**Related work.** Piecewise-linear functions and triangulated irregular networks are common objects in computational geometry and geometric information systems, and the problem of comparing two terrains defined over the same domain has been addressed in previous work [2, 13]. One focus of that work was on identifying a linear dependence between two functions or terrains. Agarwal *et al.* [2] studied the problem of computing the minimal distance  $\sigma_p(f, g) = \min_{s,t} \|sf + t - g\|_p$  for  $p = 1, \infty$ . They showed that  $\sigma_\infty(f, g)$  can be computed in  $\tilde{O}(n^{4/3})$  time (as stated in the abstract,  $\tilde{O}$  notation hides polylogarithmic factors), and that an  $(1 + \varepsilon)$ -approximation of  $\sigma_1(f, g)$  can be computed in  $O(n/\varepsilon)$  time if  $\mathbf{T}_f$  and  $\mathbf{T}_g$  are identical. However if  $\mathbf{T}_f$  and  $\mathbf{T}_g$  are different, the runtime of their algorithm is  $O(\kappa/\varepsilon)$ , where  $\kappa$  is the complexity of the overlay of the two triangulations;  $\kappa = \Theta(n^2)$  in the worst case.

Moroz and Aronov [13] showed that even for unaligned triangulations,  $\|f - g\|_2$  and  $\sigma_2(f, g)$  can be computed in  $O(n \log^4 n \log \log n)$  time. This bound also applies to computing the  $L_{2p}$ -distance, where  $p$  is a constant positive integer. Unfortunately, these approaches do not extend to computing  $\sigma_1(f, g)$  or  $\|f - g\|_1$  in subquadratic time in the general case. One notable obstacle is that the  $L_1$ -distance involves the absolute value function, which is not algebraic.

One key algebraic tool used in [13] to speed up the computation of  $\|f - g\|_2$  is the evaluation of a *univariate* polynomial of degree  $n$  on  $n$  points in  $\tilde{O}(n)$  operations. In the Real RAM model, such fast *univariate multipoint evaluation* algorithms have been known since 1972 [11]. For bivariate polynomials, it is only more recently, in 2004, that subquadratic bounds were found for their evaluation on multiple points [14]. The problem of bounding the number of operations to evaluate a multivariate polynomial on multiple points is still an active area of research, with recent results improving the best known bounds [16].

**Our results.** The main result of the paper is a subquadratic algorithm for computing  $\|f - g\|_1$ . Its runtime depends on  $\omega$ , the standard matrix multiplication constant, which is known to be less than 2.371552 [17] and conjectured by some to be 2. The bound in the following theorem holds if  $(\omega + 1)/2 > 8/5$ .

**Theorem 1.** *Let  $f$  and  $g$  be two bivariate functions defined by linear interpolation over triangulations  $\mathbf{T}_f$  and  $\mathbf{T}_g$ , respectively, over a common (bounded) domain  $\mathbb{M}$  with a total of  $n$  triangles, then  $\|f - g\|_1$  can be computed in  $\tilde{O}(n^{(\omega+1)/2})$  time, where  $\Theta(n^\omega)$  is the time required to multiply two  $n \times n$  matrices; currently the best known value of  $\omega$  is around 2.37. If the complexity of the overlay of  $\mathbf{T}_f$  and  $\mathbf{T}_g$  is  $\kappa$ , then  $\|f - g\|_1$  can be computed in  $\tilde{O}(\kappa^{(\omega-1)/2} n^{(3-\omega)/2})$  operations.*

This result is based on three main ideas: we first show that the problem can be reduced to computing a double sums of rational functions, where the denominators depend only on two variables of the outer sum (Section 2). Those sums are defined over pairs of geometric objects that satisfy specific algebraic inequalities. Next, we use geometric cuttings to obtain a sign-invariant biclique decomposition of those pairs of geometric objects (Section 3). Finally, we extend the fast multipoint polynomial evaluation (Section 4) to evaluate those sums efficiently.

Very recently, we have become aware of a preprint improving the bound on fast multipoint polynomial evaluation [16]. This seems to improve the bound of Theorem 1 to  $O(n^{8/5+\varepsilon})$  for any  $\varepsilon > 0$ . We discuss this (possible) improvement at the end of this paper.

## 2 Reduction to Multi-point Evaluation

In this section, we first express the desired quantity  $\|f - g\|_1$  as a sum of integrals associated with vertices of the overlay of  $\mathbf{T}_f$  and  $\mathbf{T}_g$  (in Section 2.1) and ultimately as a sum of low-degree rational expressions of the parameters of  $f$  and  $g$  (in Section 2.2).

Without loss of generality, we can rotate the  $(x, y)$ -coordinate system such that none of the edges of  $\mathbf{T}_f$  and  $\mathbf{T}_g$  is parallel to the vertical line  $x = 0$ . Moreover, for two triangles  $T_1 \in \mathbf{T}_f$  and  $T_2 \in \mathbf{T}_g$ , let  $\pi_1$  be the plane in  $\mathbb{R}^3$  supporting the triangle  $f(T_1)$ , and  $\pi_2$  be the plane supporting the triangle  $g(T_2)$ . If  $\pi_1$  and  $\pi_2$  are not parallel, for clarity of presentation, and with no loss of generality, we assume that their intersection line crosses the vertical plane  $x = 0$ .

### 2.1 Reduction to overlay vertices

Let  $\mathbf{V}_f, \mathbf{E}_f$  (resp.,  $\mathbf{V}_g, \mathbf{E}_g$ ) be the set of vertices and edges of  $\mathbf{T}_f$  (resp.,  $\mathbf{T}_g$ );  $\mathbf{E}_f$  and  $\mathbf{E}_g$  do not include edges contained in the boundary of  $\mathbb{M}$ . Let  $\Sigma$  be the overlay of  $\mathbf{T}_f$  and  $\mathbf{T}_g$ , a convex subdivision in  $\mathbb{R}^2$  that is a refinement of both  $\mathbf{T}_f$  and  $\mathbf{T}_g$ . Let  $\mathbf{V}$  (resp.,  $\mathbf{C}$ ) be the set of vertices (resp., bounded cells) of  $\Sigma$ , and let  $\mathbf{X} \subseteq \mathbf{E}_f \times \mathbf{E}_g$  be the family of pairs of segments that intersect. Following the same argument as in [13, Corollary 4],  $\|f - g\|_1$  can be expressed as

$$\|f - g\|_1 = \sum_{v \in \mathbf{V}} \sum_{\tau \in \mathbf{C}: \partial\tau \ni v} F(v, \tau, h_\tau), \quad (2)$$

where the function  $F(v, \tau, h_\tau)$  is defined as follows. Let  $e_1, e_2$  be the two edges of  $\tau$  incident on  $v$ , let  $\ell_i$  be the line supporting  $e_i$  for  $i = 1, 2$ , and let  $f_\tau$  (resp.  $g_\tau$ ) be the linear function of  $f$  (resp.  $g$ ) over  $\tau$ . By our assumption,  $v$  lies to the right of the  $y$ -axis. Let  $\Delta(v)$  be the triangle defined by the  $y$ -axis and the lines  $\ell_1, \ell_2$  (lying to the right of the  $y$ -axis), let  $h_\tau(x, y) = |f_\tau(x, y) - g_\tau(x, y)|$ , let  $\delta(v, \tau) \in \{-1, +1\}$  be  $-1$  if  $\tau$  lies above or below both  $\ell_1$  and  $\ell_2$ , and  $+1$  if it lies between them. See Figure 1. Then

$$F(v, \tau, h_\tau) = \delta(v, \tau) \int_{\Delta(v)} h_\tau(x, y) dx dy. \quad (3)$$

A vertex in  $\Sigma$  is either an original vertex of a  $\mathbf{T}_f$  or  $\mathbf{T}_g$  or an intersection point of an edge of  $\mathbf{T}_f$  with an edge of  $\mathbf{T}_g$ . We can rewrite (2) as:

$$\|f - g\|_1 = \sum_{v \in \mathbf{V}_f \cup \mathbf{V}_g} \sum_{\tau \in \mathbf{C}: \partial\tau \ni v} F(v, \tau, h_\tau) + \sum_{(e_1, e_2) \in \mathbf{X}} \sum_{\tau \in \mathbf{C}: \partial\tau \ni e_1 \cap e_2} F(e_1 \cap e_2, \tau, h_\tau), \quad (4)$$

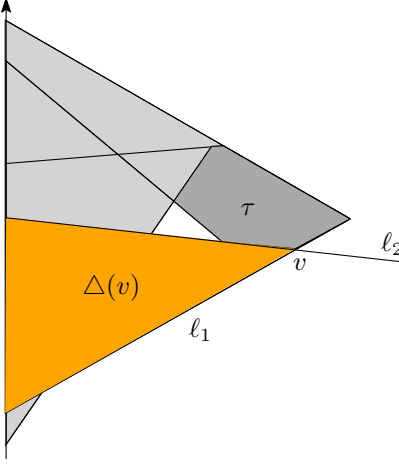


Figure 1: A typical triangle  $\Delta(v)$  associated with vertex  $v = \ell_1 \cap \ell_2$  in  $h_0$  (adapted from [13])

By preprocessing  $\mathbf{T}_f$  and  $\mathbf{T}_g$  for planar point-location queries, the cells of  $\mathbf{C}$  incident on the vertices of  $\mathbf{V}_f \cup \mathbf{V}_g$  can be computed in a total time of  $O(n \log n)$ , and thus the first term in (4) can be computed in  $O(n \log n)$  time; see [13] for details. We thus focus on computing the second term, which we denote by  $\Phi(\mathbf{E}_f, \mathbf{E}_g)$ .

For an intersecting pair of edges  $(e_1, e_2) \in \mathbf{X}$ , the vertex  $e_1 \cap e_2 \in \mathbf{V}$  is incident on four cells of  $\Sigma$ , and thus it contributes to four terms in  $\Phi(\mathbf{E}_f, \mathbf{E}_g)$ . Each of these terms has the following form: Let  $f_{11}, f_{12}$  (resp.  $g_{21}, g_{22}$ ) be the linear functions of  $f$  (resp.  $g$ ) over the triangles lying above and below  $e_1$  (resp.  $e_2$ ) respectively. For  $i, j \in \{1, 2\}$ , define  $h_{ij}(x, y) = |f_{1i}(x, y) - g_{2j}(x, y)|$ ,  $\mathbb{1}(i = j)$  to be 1 if  $i = j$  and 0 otherwise, and

$$F_{ij}(e_1, e_2) = (-1)^{\mathbb{1}(i=j)} \int_{\Delta(e_1 \cap e_2)} h_{ij}(x, y) dx dy;$$

in words,  $F_{ij}(e_1, e_2)$  is the signed volume between the graphs of  $f_{1i}$  and  $g_{2j}$  over  $\Delta(e_1 \cap e_2)$ . Using these definitions, for a pair  $(e_1, e_2) \in \mathbf{X}$ , we can write

$$\sum_{\tau \in \mathbf{C}: \partial\tau \ni e_1 \cap e_2} F(e_1 \cap e_2, \tau, h_\tau) = \sum_{i, j \in \{1, 2\}} F_{ij}(e_1, e_2) \quad (5)$$

and thus

$$\Phi(\mathbf{E}_f, \mathbf{E}_g) = \sum_{i, j \in \{1, 2\}} \sum_{(e_1, e_2) \in \mathbf{X}} F_{ij}(e_1, e_2). \quad (6)$$

In the next subsection we show that each  $F_{ij}$  can be written as the sum of at most five rational functions of constant degree each.

## 2.2 Reduction to a sum of rational expressions

We now reexamine the integrals that appear in the previous section and express them as sums of low-degree rational functions, with appropriate signs. We first focus on a single function  $F_{ij}(e_1, e_2)$ .

Let  $h_0$  be the halfplane defined by  $z = 0$  and  $x \geq 0$ . Let  $f(x, y) = a_1x + b_1y + c_1$  and  $g(x, y) = a_2x + b_2y + c_2$  be two linear functions in  $x, y$ . By abuse of notation, we let  $e_1(x) = \alpha_1x + y_1$  and  $e_2(x) = \alpha_2x + y_2$  be the two linear functions in  $x$  extending the edges  $e_1$  and  $e_2$ . Let  $h_1$  and

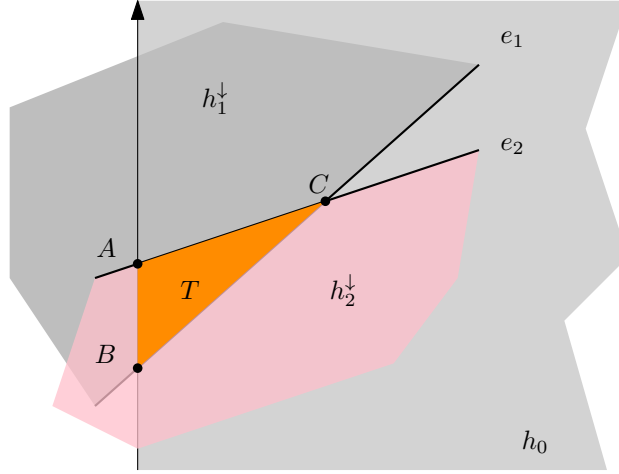


Figure 2: A typical triangle  $T = \triangle ABC$  in  $h_0$

$h_2$  be the two halfplanes in  $\mathbb{R}^3$  defined by  $z = f(x, y), y \leq e_1(x)$  and  $z = g(x, y), y \geq e_2(x)$  respectively. Below we consider such pairs  $(h_1, h_2)$  that the intersection of the projections  $h_1^\perp$  of  $h_1$  and  $h_2^\perp$  of  $h_2$  to  $h_0$  is the triangle  $T = \triangle ABC$  defined as the set of points  $(x, y)$  with  $x \geq 0$  and  $e_2(x) \leq y \leq e_1(x)$ ; refer to Figure 2.

The goal of this section is to prove that the volume between  $h_1$  and  $h_2$  over  $T$  is a sum of rational expression in the parameters of  $h_1$  and  $h_2$ , whose exact form depends on the sign variables  $\delta_A, \delta_B, \delta_C$  (i.e., values  $\pm 1$ ), indicating whether  $h_1$  is above (+1) or below (-1)  $h_2$  at the corresponding vertex of  $T$ .

More precisely, the halfplane  $h_1$  is contained in the plane  $\pi_1$  defined by  $z = f(x, y)$  and bounded by the edge  $e_1^\uparrow$ , while  $h_2$  is contained in  $\pi_2$  defined by  $z = g(x, y)$  and bounded by  $e_2^\uparrow$  such that:

$$h_1 \left| \begin{array}{l} \pi_1 : z = a_1x + b_1y + c_1 \\ e_1^\uparrow : \begin{pmatrix} t \\ y_1 + \alpha_1 t \\ z_1 + \beta_1 t \end{pmatrix} \end{array} \right. \quad h_2 \left| \begin{array}{l} \pi_2 : z = a_2x + b_2y + c_2 \\ e_2^\uparrow : \begin{pmatrix} t \\ y_2 + \alpha_2 t \\ z_2 + \beta_2 t \end{pmatrix} \end{array} \right. \quad (7)$$

Moreover, since  $e_1^\uparrow \subset \pi_1$  and  $e_2^\uparrow \subset \pi_2$ , the following relations hold:

$$\begin{cases} z_1 = b_1y_1 + c_1 \\ \beta_1 = a_1 + b_1\alpha_1 \end{cases} \quad \begin{cases} z_2 = b_2y_2 + c_2 \\ \beta_2 = a_2 + b_2\alpha_2 \end{cases} \quad (8)$$

After the same preprocessing as in [13], we can assume that  $y_1 > y_2$ ,  $\alpha_1 < \alpha_2$ , so that the projection of the edges intersect at  $x = \frac{y_1 - y_2}{\alpha_2 - \alpha_1} > 0$ . Thus, on the  $xy$ -plane, the triangle  $T$  has three vertices  $A, B, C$ :

$$A \left| \begin{array}{l} 0 \\ y_1 \end{array} \right. \quad B \left| \begin{array}{l} 0 \\ y_2 \end{array} \right. \quad C \left| \begin{array}{l} \frac{y_1 - y_2}{\alpha_2 - \alpha_1} \\ \frac{\alpha_2 y_1 - \alpha_1 y_2}{\alpha_2 - \alpha_1} \end{array} \right. \quad (9)$$

### 2.2.1 Explicit sign conditions

The explicit rational expressions for  $\int_T |f - g|$  will depend on the value of  $\delta_A, \delta_B, \delta_C$ , namely the sign of the difference  $g - f$  above  $A, B$ , and  $C$  respectively. Combining eqs. (7), (8), and (9) and

using the assumption  $\alpha_2 > \alpha_1$ , we can explicitly expand the formula for  $\delta_A$ ,  $\delta_B$ , and  $\delta_C$ :

$$\begin{aligned}
\delta_A &= \text{sign}(b_2 y_1 + c_2 - z_1) \\
\delta_B &= \text{sign}(z_2 - b_1 y_2 - c_1) \\
\delta_C &= \text{sign}\left(\frac{(z_2 - z_1)(\alpha_2 - \alpha_1) + (\beta_2 - \beta_1)(y_1 - y_2)}{\alpha_2 - \alpha_1}\right) \\
&= \text{sign}((z_2 - z_1)(\alpha_2 - \alpha_1) + (\beta_2 - \beta_1)(y_1 - y_2)) \\
&= \text{sign}(\gamma_2 - \gamma_1 + \beta_2 y_1 - z_1 \alpha_2 + \beta_1 y_2 - z_2 \alpha_1)
\end{aligned} \tag{10}$$

where  $\gamma_i = z_i \alpha_i - \beta_i y_i$ , and  $\text{sign}(x) = +1$  if  $x \geq 0$  and  $-1$  otherwise. Note that, after the introduction of  $\gamma_i$ , the formula for  $\delta_A, \delta_B, \delta_C$  are the signs of three *bilinear* expressions in the five parameters  $\alpha_i, \beta_i, y_i, z_i, \gamma_i$  describing each of the two halfplanes.

### 2.2.2 Explicit rational expressions

The volume  $\int_T |f - g|$  will be written as a weighted sum of three rational expressions. For that, we introduce the intersection point  $D$  of the line  $e_1^\uparrow$  with the plane  $\pi_2$ , the intersection point  $E$  of  $e_2^\uparrow$  and  $\pi_1$ , and the intersection point  $F$  of the line  $DE$  with the vertical plane  $x = 0$ . These points exist due to our general position assumptions, as soon as  $\delta_A, \delta_B$  and  $\delta_C$  don't all have the same sign.

Let  $D^\downarrow, E^\downarrow$ , and  $F^\downarrow$  be projections of  $D, E$ , and  $F$  respectively to  $h_0$ . Finally, the points  $A_1, B_1, C_1$  are the points above  $A, B, C$  respectively, with  $z$ -coordinates  $f(x_A, y_A), f(x_B, y_B)$  and  $f(x_C, y_C)$ , and the points  $A_2, B_2, C_2$  are the points above  $A, B, C$  respectively, with  $z$ -coordinates  $g(x_A, y_A), g(x_B, y_B)$  and  $g(x_C, y_C)$ . Figures 3 and 4 illustrate those notations in different cases.

We will express the volume between  $h_1$  and  $h_2$  over  $T$  as a weighted sum of volumes of polytopes spawned by  $C, D$ , and  $E$  respectively, toward the vertical plane  $x = 0$ . The following expressions represent the signed volumes of the polytopes between the planes  $\pi_1$  and  $\pi_2$ , above the triangles  $T = \triangle ABC, \triangle AD^\downarrow F^\downarrow$ , and  $\triangle BE^\downarrow F^\downarrow$  respectively. More precisely, letting  $R_C = \int_T (g - f)$ ,  $R_D = \int_{\triangle AD^\downarrow F^\downarrow} (g - f)$  and  $R_E = \int_{\triangle BE^\downarrow F^\downarrow} (g - f)$ , we have:

$$\begin{aligned}
R_C &= \frac{1}{6} x_C (y_A - y_B) \sum_{X \in \{A, B, C\}} (g(x_X, y_X) - f(x_X, y_X)), \\
R_D &= \frac{1}{6} x_D (y_F - y_A) (g(x_A, y_A) - f(x_A, y_A)), \text{ and} \\
R_E &= \frac{1}{6} x_E (y_F - y_B) (g(x_B, y_B) - f(x_B, y_B)).
\end{aligned} \tag{11}$$

We now proceed to express  $R_C, R_D$ , and  $R_E$  as fractions in the parameters of  $h_1$  and  $h_2$ . In particular, we will show that the denominator of each fraction depends only on either two parameters of  $h_1$ , or two parameters of  $h_2$ .

**Rational representation of  $R_C$**  Using the formula for the signed volume  $R_C$  in eq. (11), we expand the evaluation of  $f$  and  $g$  at  $A, B, C$  to obtain

$$R_C = \frac{1}{6}(y_1 - y_2) \frac{y_1 - y_2}{\alpha_2 - \alpha_1} \left( \left\langle \begin{pmatrix} a_1 - a_2 \\ b_1 - b_2 \end{pmatrix}, A + B + C \right\rangle + 3(c_1 - c_2) \right) \\ = \frac{(y_1 - y_2)^2 \left[ (\alpha_2 - \alpha_1) [(b_1 - b_2)(y_1 + y_2) + 3(c_1 - c_2)] + (a_1 - a_2)(y_1 - y_2) + (b_1 - b_2)(\alpha_2 y_1 - \alpha_1 y_2) \right]}{6(\alpha_1 - \alpha_2)^2}, \quad (12)$$

so that  $R_C$  is a rational expression with denominator  $(\alpha_2 - \alpha_1)^2$ .

**Rational representation of  $R_D$  and  $R_E$**  For quantities  $R_D$  and  $R_E$ , we need to compute the coordinates of  $D$ ,  $E$ , and  $F$ . Using eq. (7), the abscissa  $t_D$  of  $D$  satisfies  $z_1 + \beta_1 t_D = a_2 t_D + b_2 y_1 + b_2 \alpha_1 t_D + c_2$ . A similar relation for the abscissa  $t_E$  of  $E$  leads to

$$t_D = \frac{z_1 - b_2 y_1 - c_2}{b_2 \alpha_1 + a_2 - \beta_1} \quad t_E = \frac{z_2 - b_1 y_2 - c_1}{b_1 \alpha_2 + a_1 - \beta_2} \\ = \frac{(b_1 - b_2) y_1 + c_1 - c_2}{(b_2 - b_1) \alpha_1 + a_2 - a_1} \quad = \frac{(b_1 - b_2) y_2 + c_1 - c_2}{(b_2 - b_1) \alpha_2 + a_2 - a_1}$$

Moreover,  $F$  is at the intersection of  $h_1$ ,  $h_2$ , and the vertical plane  $x = 0$ . In particular, its coordinates are the solution of  $x = 0, z = b_1 y + c_1 = b_2 y + c_2$ , so we have

$$D \left| \begin{array}{c} t_D \\ y_1 + \alpha_1 t_D \\ z_1 + \beta_1 t_D \end{array} \right. \quad E \left| \begin{array}{c} t_E \\ y_2 + \alpha_2 t_E \\ z_2 + \beta_2 t_E \end{array} \right. \quad F \left| \begin{array}{c} 0 \\ \frac{c_2 - c_1}{b_1 - b_2} \\ \frac{b_1 c_2 - b_2 c_1}{b_1 - b_2} \end{array} \right. \quad (13)$$

Also, the points  $A_1, A_2, B_1, B_2$  have coordinates

$$A_1 \left| \begin{array}{c} 0 \\ y_1 \\ b_1 y_1 + c_1 \end{array} \right. \quad A_2 \left| \begin{array}{c} 0 \\ y_1 \\ b_2 y_1 + c_2 \end{array} \right. \quad B_1 \left| \begin{array}{c} 0 \\ y_2 \\ b_1 y_2 + c_1 \end{array} \right. \quad B_2 \left| \begin{array}{c} 0 \\ y_2 \\ b_2 y_2 + c_2 \end{array} \right. \quad (14)$$

Finally, replacing the coordinates in eq. (11), this leads to:

$$R_D = \frac{1}{6} \frac{(b_1 - b_2) y_1 + c_1 - c_2}{(b_2 - b_1) \alpha_1 + a_2 - a_1} \left( \frac{c_2 - c_1}{b_1 - b_2} - y_1 \right) (b_2 y_1 + c_2 - b_1 y_1 - c_1) \\ = \frac{1}{6} \frac{((b_1 - b_2) y_1 + c_1 - c_2)^3}{(b_2 - b_1) \alpha_1 + a_2 - a_1} \quad (15)$$

and

$$R_E = \frac{1}{6} \frac{(b_1 - b_2) y_2 + c_1 - c_2}{(b_2 - b_1) \alpha_2 + a_2 - a_1} \left( \frac{c_2 - c_1}{b_1 - b_2} - y_2 \right) (b_2 y_2 + c_2 - b_1 y_2 - c_1) \\ = \frac{1}{6} \frac{((b_1 - b_2) y_2 + c_1 - c_2)^3}{(b_2 - b_1) \alpha_2 + a_2 - a_1}. \quad (16)$$

### 2.2.3 Case distinctions for the volume expression

Our goal is to provide a formula for the volume of the polyhedron between  $\pi_1$  and  $\pi_2$  above the triangle  $ABC$ , as a linear combination of fractions where the denominators depend only on two variables from the triangulation  $\mathbf{T}_f$  or on two variables from  $\mathbf{T}_g$ . This will allow us to use fast multipoint evaluation algorithms for bivariate polynomials (see Section 4).

We will see that for the different cases of signs of  $\delta_A, \delta_B, \delta_C$ , the volume above  $T$  can be expressed as a weighted signed sum of  $R_C, R_D$ , and  $R_E$ . We consider the cases where the vector of signs  $(\delta_A, \delta_B, \delta_C)$  is  $(+1, +1, +1), (+1, +1, -1)$ , or  $(+1, -1, -1)$ . The other cases are deduced by symmetry.

**When  $\delta_A, \delta_B, \delta_C$  are positive: the halfplanes don't meet above the triangle.** In this case,  $V_{+++}(h_1, h_2) = R_C$ , a rational expression with denominator  $(\alpha_2 - \alpha_1)^2$ . By symmetry, when  $\delta_A, \delta_B, \delta_C$  are  $-1$ , we have  $V_{---}(h_1, h_2) = -R_C$ .

**When  $\delta_A, \delta_B$  are positive and  $\delta_C$  is negative.** Using the notation of Section 2.2.2, the volume between the planes over  $T$  in this case is the sum of the volumes  $\text{vol}(A_1A_2B_2B_1DE)$  and  $\text{vol}(DEC_2C_1)$ ; refer to Figure 3. With  $\delta_A, \delta_B$  positive, and  $\delta_C$  negative, the signed volume  $R_C$  is equal to  $\text{vol}(A_1A_2B_1DE) - \text{vol}(DEC_2C_1)$ . Thus, the volume  $V_{++-}$  can be expressed as

$$V_{++-}(h_1, h_2) = 2 \text{vol}(A_1A_2B_2B_1DE) - R_C.$$

The volume of the polytope  $A_1A_2B_2B_1DE$  is computed as the difference of volumes of the two simplices  $B_1B_2FE$  and  $A_1A_2FD$ . When  $\delta_A$  and  $\delta_B$  are positive, the  $y$ -coordinate of the point  $F$  can be either (i) greater than  $y_A$ , or (ii) less than  $y_B$ .

In case (i), we have  $\text{vol}(A_1A_2B_2B_1DE) = \text{vol}(B_1B_2FE) - \text{vol}(A_1A_2FD)$ . Moreover  $\text{vol}(A_1A_2FD) = R_D$  and  $\text{vol}(B_1B_2FE) = R_E$ , so that  $\text{vol}(A_1A_2B_2B_1DE) = R_E - R_D$ .

In case (ii), we have  $\text{vol}(A_1A_2B_2B_1DE) = \text{vol}(A_1A_2FD) - \text{vol}(B_1B_2FE)$ ,  $\text{vol}(A_1A_2FD) = -R_D$ , and  $\text{vol}(B_1B_2FE) = -R_E$ , so that once again  $\text{vol}(A_1A_2B_2B_1DE) = R_E - R_D$ .

Thus, in both cases, we have  $\text{vol}(A_1A_2B_2B_1DE) = R_E - R_D$ , and therefore

$$V_{++-}(h_1, h_2) = 2R_E - 2R_D - R_C.$$

Similarly, by symmetry, we have  $V_{--+} = -2R_E + 2R_D + R_C$ .

**When  $\delta_A$  is positive and  $\delta_B, \delta_C$  are negative.** This case is illustrated in Figure 4. Using the notation of the previous section the volume between the two halfplanes is the sum of volumes of the polytopes  $A_1A_2FD$  and  $B_1B_2C_2C_1DF$ . First, with  $\delta_A$  positive and  $\delta_B, \delta_C$  negative, the signed volume  $R_C$  is  $\text{vol}(A_1A_2FD) - \text{vol}(B_1B_2C_2C_1DF)$ . Moreover, since  $\delta_A$  is positive and  $\delta_B$  is negative, the  $y$ -coordinate of  $F$  is in the interval  $[y_B, y_A]$  and we have  $\text{vol}(A_1A_2FD) = -R_D$ . Thus, the volume  $V_{+--}$  can be expressed as

$$V_{+--}(h_1, h_2) = 2 \text{vol}(A_1A_2FD) - R_C = -2R_D - R_C. \quad (17)$$

By symmetry, we have  $V_{-++} = 2R_D + R_C$ . Moreover, the role of  $D$  and  $E$  are symmetric, and when  $\delta_A, \delta_C$  are negative, and  $\delta_B$  is positive, we have similarly  $V_{-+-} = 2 \text{vol}(B_2B_1FE) - R_C = 2R_E - R_C$  and  $V_{+-+} = -2R_E + R_C$ .

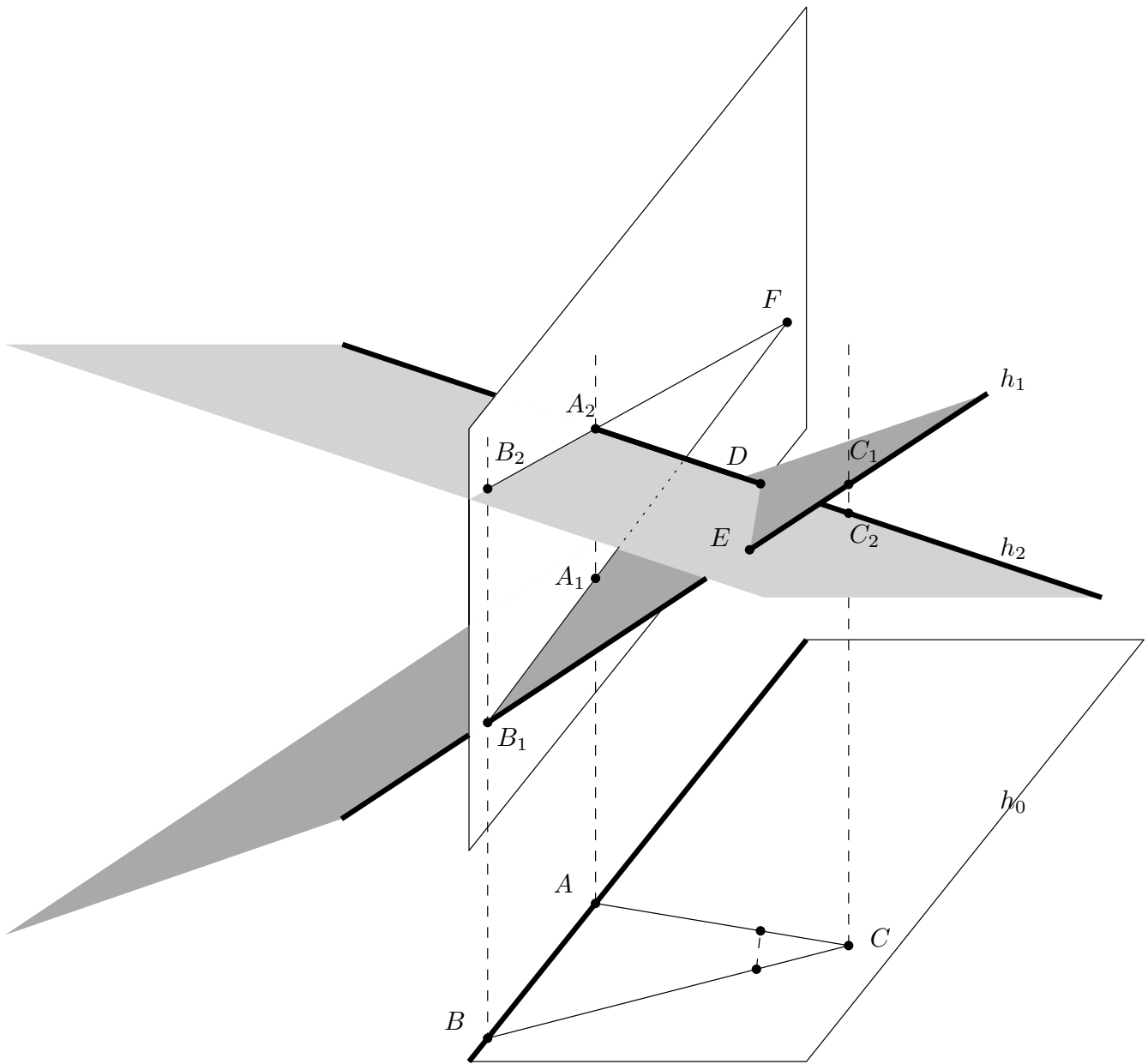


Figure 3: Case where  $\delta_C$  has a sign different from  $\delta_A$  and  $\delta_B$

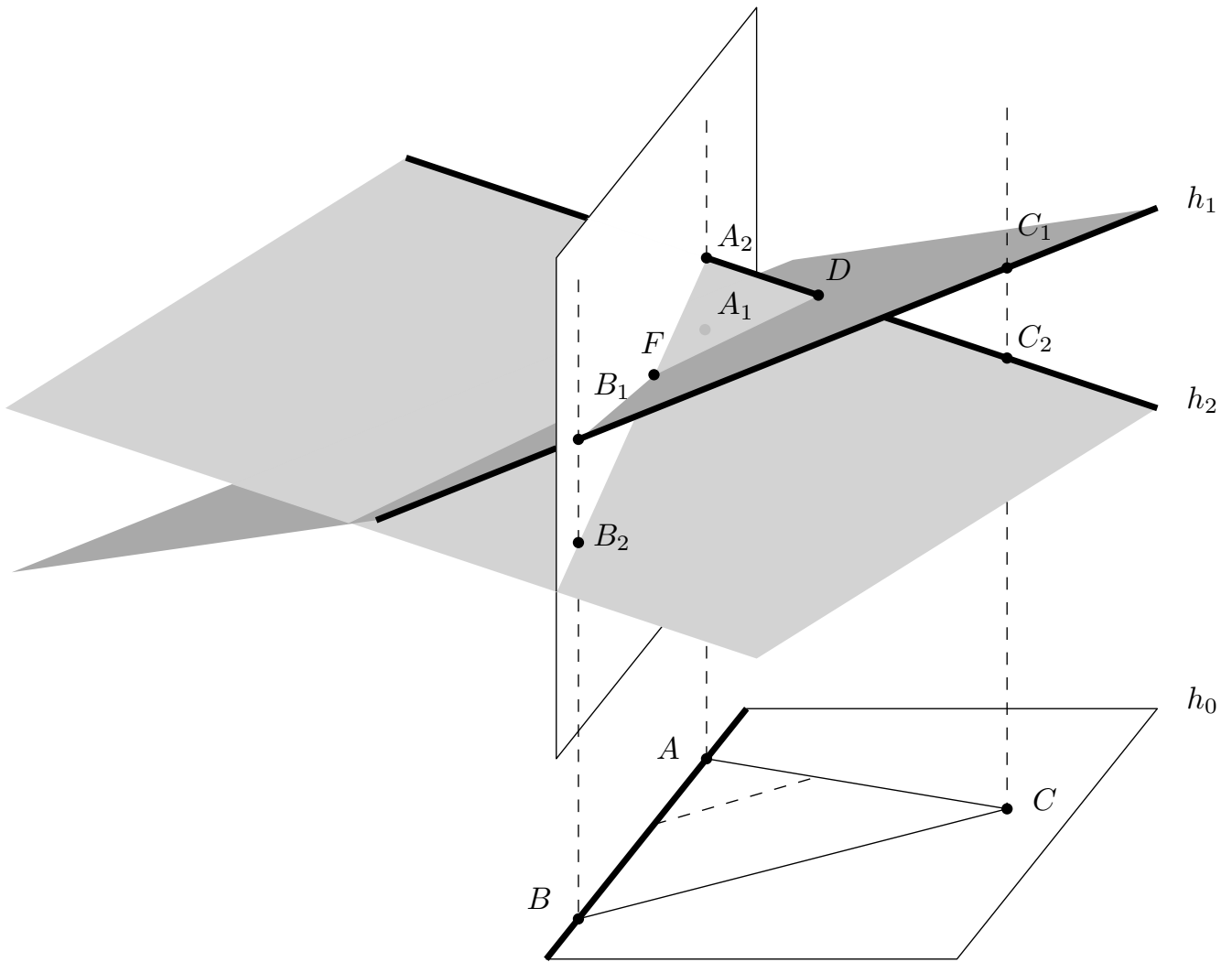


Figure 4: Case where  $\delta_A$  has a sign different from  $\delta_B$  and  $\delta_C$

**Gathering all the cases.** Depending on the signs  $\delta_A$ ,  $\delta_B$ , and  $\delta_C$ , we have different formulas for the volume between the planes over  $T$ . We can gather them all within one expression that is bilinear in  $\delta_A, \delta_B, \delta_C$  on one side, and  $R_A, R_B, R_C$  on the other side.

**Lemma 2.** Assume that  $v_1 = (a_1, b_1, c_1, y_1, \alpha_1)$  and  $v_2 = (a_2, b_2, c_2, y_2, \alpha_2)$  are vectors of real numbers such that  $y_1 > y_2$  and  $\alpha_1 < \alpha_2$ . Let  $T \subset \mathbb{R}^2$  be a triangular cell defined by the set of points  $(x, y)$  such that  $x \geq 0$ ,  $y \leq y_1 + \alpha_1 x$ , and  $y \geq y_2 + \alpha_2 x$ . And let  $f$  and  $g$  be two linear functions defined by  $f(x, y) = a_1 x + b_1 y + c_1$  and  $g(x, y) = a_2 x + b_2 y + c_2$ . Then the volume  $\int_{(x,y) \in T} |f - g| dx dy$  can be expressed as:

$$\int_{(x,y) \in T} |f - g| dx dy = \delta_C(R_D - R_E + R_C) - \delta_A R_D + \delta_B R_E$$

where  $\delta_A, \delta_B, \delta_C$ , defined in eq. (10), depend on at most four entries of  $v_1$  and four entries of  $v_2$ , and:

$$R_C(v_1; v_2) = \frac{p_C(v_1; v_2)}{q_C(\alpha_1; \alpha_2)} \quad R_D(v_1; v_2) = \frac{p_A(v_1, v_2)}{q_A(v_1; (a_2, b_2))} \quad R_E(v_1; v_2) = \frac{p_B(v_1, v_2)}{q_B((a_1, b_1); v_2)}$$

are rational functions of constant degrees defined in eqs. (12), (15) and (16).

**Remark.** If instead of the assumptions  $y_1 > y_2$  and  $\alpha_1 < \alpha_2$ , we have the assumptions  $y_2 < y_1$  and  $\alpha_2 < \alpha_1$ , then the signed volume above the triangle  $ABC$  becomes  $-R_C$ , and  $\delta_C$  becomes  $-\delta_C$  since we factored out the sign of  $\alpha_2 - \alpha_1$  in eq. (10). Without assumption on the order between  $\alpha_1$  and  $\alpha_2$ , letting  $\sigma = \text{sign}(\alpha_2 - \alpha_1)$ , the formula for the volume becomes:

$$\int_{(x,y) \in T} |f - g| dx dy = \delta_C R_C + \delta_C \sigma (R_D - R_E) - \delta_A R_D + \delta_B R_E.$$

Returning to expressing (6) as a sum of rational functions, recall that for  $j \in 1, 2$ ,

$$F_{ij}(e_1, e_2) = (-1)^{\mathbb{1}(i=j)} \left[ \delta_C(e_1^*, e_2^*) R_C(e_1^*, e_2^*) + \sigma(e_1^*, e_2^*) \delta_C(e_1^*, e_2^*) (R_D(e_1^*, e_2^*) - R_E(e_1^*, e_2^*)) - \delta_A(e_1^*, e_2^*) R_D(e_1^*, e_2^*) + \delta_B(e_1^*, e_2^*) R_E(e_1^*, e_2^*) \right].$$

Since there are five terms in the above expression for  $F_{ij}$ , eq. (6) can be expressed as follows:

**Corollary 3.** Let  $E_f, E_g$  be the set of edges in  $\mathbf{T}_f$  and  $\mathbf{T}_g$  respectively, and  $X \subseteq E_f \times E_g$  be the intersecting pairs of segments. Then

$$\Phi(E_f, E_g) = \sum_{k=1}^5 \sum_{i,j \in \{1,2\}} \sum_{(e_1, e_2) \in X} (-1)^{\mathbb{1}(i=j)} \sigma_{ijk}(e_{1i}^*, e_{2j}^*) \cdot \xi_{ijk}(e_{1i}^*, e_{2j}^*) \cdot F_{ijk}(e_{1i}^*, e_{2j}^*), \quad (18)$$

where  $\sigma_{ijk}$  is either  $\sigma$  or the constant function 1,  $\xi_{ijk} \in \{\pm\delta_A, \pm\delta_B, \pm\delta_C\}$ , and  $F_{ijk} \in \{R_C, R_D, R_E\}$ .

### 3 Computing a Biclique Partition

Using Corollary 3,  $\Phi(\mathbf{E}_f, \mathbf{E}_g)$  can be computed in  $O(\kappa)$  time, where  $\kappa := |\mathbf{X}|$  is the number of intersecting pairs of edges in  $\mathbf{E}_f \times \mathbf{E}_g$ , in a straightforward manner if we have the overlay  $\Sigma$  of  $\mathbf{T}_f$  and  $\mathbf{T}_g$  at our disposal. In this section, we show how to “batch” the evaluations of  $\Phi_{ijk}$ , using biclique partitions of  $\mathbf{X}$  so that the total time spent is  $\tilde{O}(\kappa^{(\omega-1)/2} n^{(3-\omega)/2})$ . Recall that for an edge  $e_i \in \mathbf{E}_f \cup \mathbf{E}_g$ ,  $e_{i1}^*$  (resp.  $e_{i2}^*$ ) is the point in  $\mathbb{R}^5$  corresponding to the linear function over the triangle lying below (resp. above) the line supporting  $e_i$ . We describe the algorithm for computing

$$\Phi_C(\mathbf{E}_f, \mathbf{E}_g) = \sum_{(e_1^*, e_2^*) \in \mathbf{X}} \sigma(e_1, e_2) \delta_C(e_{11}^*, e_{22}^*) R_C(e_{11}^*, e_{22}^*).$$

The other terms (those involving  $R_D$  and  $R_E$ ) can be evaluated in an analogous manner. For simplicity, we use  $e_i^*$  to denote  $e_{i1}^*$  (resp.  $e_{i2}^*$ ) if  $e_i$  is a segment of  $\mathbf{E}_f$  (resp.  $\mathbf{E}_g$ ). We compute the desired biclique partition of  $\mathbf{X}$  in two stages. The first stage is similar to that described in [13], but the second stage is more involved and relies on geometric cuttings [9, 3].

**First stage.** As in Chazelle *et al.* [10] (see also [13]), we compute, in time  $O(n \log^2 n)$ , a *biclique partition*  $\mathcal{B} = \{(R_1, B_1), \dots, (R_s, B_s)\}$  of  $\mathbf{X}$  such that

- (i) for every  $k \leq s$ ,  $R_k \subseteq \mathbf{E}_f$  and  $B_k \subseteq \mathbf{E}_g$  and every segment in  $R_k$  intersects every segment in  $B_k$ ;
- (ii) every segment in  $R_k$  has lower slope than that of every segment in  $B_k$ , or vice-versa;
- (iii) for every intersecting pair  $(e_1, e_2) \in \mathbf{X}$ , there exists exactly one  $k \leq s$  with  $e_1 \in R_k$  and  $e_2 \in B_k$ ; and
- (iv)  $\sum_k (|R_k| + |B_k|) = O(n \log^2 n)$  and  $\sum_k |R_k| \times |B_k| = \kappa$ .

For any  $1 \leq k \leq s$ ,  $\sigma(e_1^*, e_2^*)$  is the same for all  $(e_1, e_2) \in R_k \times B_k$ , which we denote by  $\sigma(R_k, B_k)$ . Therefore, we express  $\Phi_C(\mathbf{E}_f, \mathbf{E}_g)$  in terms of  $\mathcal{B}$  as follows:

$$\Phi_C(\mathbf{E}_f, \mathbf{E}_g) = \sum_{k=1}^s \sigma(R_k, B_k) \sum_{(e_1, e_2) \in R_k \times B_k} \delta_C(e_1^*, e_2^*) \cdot R_C(e_1^*, e_2^*). \quad (19)$$

We fix a biclique  $(R_k, B_k) \in \mathcal{B}$ . Without loss of generality, assume that  $\sigma(R_k, B_k) = 1$ . We describe how to compute

$$\Phi_C(R_k, B_k) = \sum_{(e_1, e_2) \in R_k \times B_k} \delta_C(e_1^*, e_2^*) \cdot R_C(e_1^*, e_2^*).$$

**Second stage.** We thus have the following problem at hand. Let  $\mathbf{R} \subseteq \mathbf{E}_f$  and  $\mathbf{B} \subseteq \mathbf{E}_g$  be two sets of segments in  $\mathbb{R}^2$  such that every pair of segments in  $\mathbf{R} \times \mathbf{B}$  intersects and the slope of every segment in  $\mathbf{R}$  is less than that of any segment in  $\mathbf{B}$ ; set  $m = |\mathbf{R}|$  and  $n = |\mathbf{B}|$ . Without loss of generality assume that  $m \leq n$ . We compute a *sign-invariant* (biclique) partition  $\Pi = \{(R_1, B_1), \dots, (R_u, B_u)\}$  of  $\mathbf{R} \times \mathbf{B}$ , as described below, such that for all  $1 \leq k \leq s$ , the value of  $\delta_C(e_i^*, e_j^*)$  is the same for all pairs  $(e_i, e_j) \in (R_k, B_k)$ , which we denote by  $\delta_C(R_k, B_k)$ . Let  $\Pi^+ = \{(R_k, B_k) \mid \delta_C(R_k, B_k) = 1\}$  and  $\Pi^- = \Pi \setminus \Pi^+$ . Then

$$\Phi_C(\mathbf{R}, \mathbf{B}) = \sum_{(R_k, B_k) \in \Pi^+} \sum_{(e_i, e_j) \in R_k \times B_k} R_C(e_i^*, e_j^*) - \sum_{(R_k, B_k) \in \Pi^-} \sum_{(e_i, e_j) \in R_k \times B_k} R_C(e_i^*, e_j^*). \quad (20)$$

We show below in Section 4 how  $\Phi_C(\mathbf{R}_k, \mathbf{B}_k) = \sum_{(e_i, e_j) \in \mathbf{R}_k \times \mathbf{B}_k} \Phi_C(e_i^*, e_j^*)$  can be computed efficiently. It thus suffices to describe how we compute  $\Pi$  and to analyze the overall run time.

By eq. (10),  $\delta_C$  is the sign function of a bilinear function  $P(\varphi, \psi)$ , where  $\varphi$  (resp.  $\psi$ ) is a set of five variables representing the parameters  $y_i, z_i, \alpha_i, \beta_i, \gamma_i$  corresponding to the edges  $e_i$  of  $\mathbf{R}$  (resp.  $\mathbf{B}$ ) used in the definition of  $\delta_C$ ; note that four of these parameters are independent and  $\gamma_i = z_i \alpha_i - \beta_i y_i$  depends on them. Abusing the notation slightly, we continue to represent an edge  $e_j \in \mathbf{R}$  as a point  $e_j^*$  in  $\mathbb{R}^5$  with the new set of parameters. We also map  $e_j$  to a hyperplane  $\tilde{e}_j$  in  $\mathbb{R}^5$ , where  $\tilde{e}_j: P(e_j^*, \psi) = 0$ . Similarly, we map an edge  $e_k \in \mathbf{B}$  to a point  $e_k^* \in \mathbb{R}^5$  and a hyperplane  $\tilde{e}_k: P(\varphi, e_k^*) = 0$  in  $\mathbb{R}^5$ . For a subset  $\mathbf{A}$  of  $\mathbf{R}$  (or  $\mathbf{B}$ ), we define  $\mathbf{A}^* = \{e_j^* \mid e_j \in \mathbf{A}\}$  and  $\tilde{\mathbf{A}} = \{\tilde{e}_j \mid e_j \in \mathbf{A}\}$ . Our goal is to compute a partition  $\Pi$  of  $\mathbf{R} \times \mathbf{B}$  such that for any pair  $(\mathbf{R}_i, \mathbf{B}_i) \in \Pi$ , all points of  $\mathbf{B}_i^*$  lie on the same side of all hyperplanes in  $\tilde{\mathbf{R}}_i$ , or equivalently all points of  $\mathbf{R}_i^*$  lie on the same side of all hyperplanes in  $\tilde{\mathbf{B}}_i$ .

Let  $\mathbf{H}$  be a set of  $m$  hyperplanes in  $\mathbb{R}^d$ ,  $\Delta$  be a simplex, and  $\chi$  be the number of vertices of the arrangement  $\mathcal{A}(\mathbf{H})$  of  $\mathbf{H}$  inside  $\Delta$ . For a parameter  $r > 1$ , a partition of  $\Delta$  into a family  $\Xi$  of simplices, referred to as *cells*, is called a  $(1/r)$ -cutting of  $\mathbf{H}$  within  $\Delta$  if every cell of  $\Xi$  is crossed by at most  $m/r$  hyperplanes of  $\mathbf{H}$ . (For  $r > m$ , cells of  $\Xi$  are not crossed by any hyperplane of  $\mathbf{H}$ , i.e.,  $\Xi$  is a refinement of  $\mathcal{A}(\mathbf{H})$ .) The *conflict list* of a cell  $\Delta \in \Xi$ , denoted by  $\mathbf{H}_\Delta$ , is the subset of hyperplanes that cross  $\Delta$ . It is well known that a  $(1/r)$ -cutting of  $\mathbf{H}$  of size  $O(r^d)$  along with the conflict lists of all of its cells can be computed in  $O(mr^{d-1})$  time [9, 12]. For any point set  $P \subset \mathbb{R}^d$ , one can compute  $P \cap \Delta$  for every  $\Delta \in \Xi$  in additional  $O(|P| \log r)$  time.

In our context, we compute the desired sign-invariant partition  $\Pi$  of  $\mathbf{R} \times \mathbf{B}$  using geometric cuttings in a round-robin manner as in [4, 12]. We choose  $r$  to be a sufficiently large constant, see below. We assume that the input is *balanced*, i.e.,  $n/2 \leq m \leq n$ . Otherwise (i.e.  $m < n/2$ ), we use the standard batching technique—we partition  $\mathbf{B}$  into  $\ell = \lceil n/m \rceil$  subsets  $\mathbf{B}_1, \dots, \mathbf{B}_\ell$ , each of size between  $m$  and  $2m$ , and compute the desired partition for each pair  $(\mathbf{R}, \mathbf{B}_j)$ ,  $1 \leq j \leq \ell$ , independently.

Each round of the algorithm works in two stages. In the first stage, we map  $\mathbf{R}$  to the set  $\tilde{\mathbf{R}}$  of hyperplanes in  $\mathbb{R}^5$  and compute a  $(1/r)$ -cutting  $\Xi$  of size  $O(r^5)$ . For each cell  $\Delta \in \Xi$ , let  $\tilde{\mathbf{R}}_\Delta \subseteq \tilde{\mathbf{R}}$  be the conflict list of  $\Delta$ , let  $\tilde{\mathbf{R}}_\Delta^+$  (resp.  $\tilde{\mathbf{R}}_\Delta^-$ ) be the set of hyperplanes of  $\tilde{\mathbf{R}}$  that lie above (resp., below)  $\Delta$ . Let  $\mathbf{R}_\Delta, \mathbf{R}_\Delta^+, \mathbf{R}_\Delta^- \subseteq \mathbf{R}$  be the sets of segments corresponding to  $\tilde{\mathbf{R}}_\Delta, \tilde{\mathbf{R}}_\Delta^+, \tilde{\mathbf{R}}_\Delta^-$ , respectively. Let  $\mathbf{B}_\Delta^* = \mathbf{B}^* \cap \Delta$  and  $\mathbf{B}_\Delta = \{e_j \mid e_j^* \in \mathbf{B}_\Delta^*\}$ . Set  $m_\Delta = |\mathbf{R}_\Delta|$  and  $n_\Delta = |\mathbf{B}_\Delta|$ . If  $n_\Delta \geq n/r^5$  for any cell  $\Delta$ , we partition  $\Delta$  further into subcells each of which contains at most  $n/r^5$  points of  $\mathbf{B}^*$ . The number of cells even after this refinement is at most  $c_0 r^5$  for some constant  $c_0 > 0$ . For every cell  $\Delta \in \Xi_i$ , we add  $(\mathbf{R}_\Delta^+, \mathbf{B}_\Delta)$  and  $(\mathbf{R}_\Delta^-, \mathbf{B}_\Delta)$  as bicliques to the partition  $\Pi$ . This completes the description of the first stage, which takes  $O(mr^4 + n \log r) = O(n)$  since  $r = O(1)$  and  $m \leq n$ .

In the second stage, for each cell  $\Delta \in \Xi$ , we map  $\mathbf{B}_\Delta$  to the set  $\tilde{\mathbf{B}}_\Delta$  of hyperplanes and compute a  $(1/r)$  cutting  $\Xi_\Delta$  of  $\tilde{\mathbf{B}}_\Delta$  of size  $O(r^5)$ . For a cell  $\tau \in \Xi_\Delta$ , we define  $\mathbf{B}_\tau, \mathbf{B}_\tau^+, \mathbf{B}_\tau^-, \mathbf{R}_\Delta$  as in the first stage (but reversing the roles of  $\mathbf{R}$  and  $\mathbf{B}$ ). If  $|\mathbf{R}_\tau| \leq m/r^6$ , then we split  $\tau$  into subcells each of which contains at most  $m/r^6$  points of  $\mathbf{R}_\tau$ . Let  $m_\tau = |\mathbf{R}_\tau|$  (after the refinement) and  $n_\tau = |\mathbf{B}_\tau|$ . By construction,  $m_\tau \leq m/r^6$  and  $n_\tau \leq n_\Delta/r \leq n/r^6$  for all  $\tau \in \Xi_\Delta$  and for all  $\Delta \in \Xi$ . For every  $\Delta \in \Xi$  and  $\tau \in \Xi_\Delta$ , we add  $(\mathbf{R}_\tau, \mathbf{B}_\tau^+)$  and  $(\mathbf{R}_\tau, \mathbf{B}_\tau^-)$  as bicliques to the partition  $\Pi$ .

For every cell  $\tau$  of a second-stage cutting, we recursively compute a partition of  $(\mathbf{R}_\tau, \mathbf{B}_\tau)$ . The recursion stops when  $\min\{|\mathbf{R}_\tau|, |\mathbf{B}_\tau|\} < r$ , in which case we add each pair  $(e_i, e_j) \in \mathbf{R}_\tau \times \mathbf{B}_\tau$  as a singleton biclique to  $\Pi$ . This completes the description of computing  $\Pi$ . The correctness follows using a standard argument, see, e.g. [3].

Finally, for each biclique  $(\mathbf{R}_k, \mathbf{B}_k) \in \Pi$ , we compute  $\Phi_C(\mathbf{R}_k, \mathbf{B}_k)$  using Corollary 7. This com-

pletes the description of the algorithm.

**Analysis** We now analyze the overall run time of the algorithm, including the time spent in computing  $\Phi_C(\mathbf{R}_k, \mathbf{B}_k)$  for each clique  $(\mathbf{R}_k, \mathbf{B}_k) \in \Pi$ . First, we focus on the run time for the balanced case. For notational simplicity, assume that  $|\mathbf{R}| = |\mathbf{B}| = n$ . Let  $T(n)$  be the total time spent in computing the partition  $\Pi$  plus the time spent in computing  $\Phi_C(\cdot, \cdot)$  for each biclique in  $\Pi$ .

If  $n < n_0$ , where  $n_0$  is a constant that depends on  $r$ , the algorithm takes  $O(1)$  time. Otherwise, the algorithm spends  $O(n)$  time in the two stages of the first round to compute the cuttings, the conflict lists, and the bicliques. Then it solves at most  $c_1 r^{10}$  recursive problems, where  $c_1 > 0$  is a constant independent of  $r$ , each of size at most  $n/r^6$ . Furthermore, the first round generates at most  $O(r^{10})$  bicliques, and by Corollary 7, computing  $\Phi_C$  for each biclique takes  $\tilde{O}(n^{(\omega+1)/2})$ . Thus the total time spent in the first round in computing the bicliques and  $\Phi_C$  for all these is at most  $c_3 n^{(\omega+1)/2} \log^{c_2} n$ , where  $c_2 > 0$  is a constant independent on  $r$  and  $c_3 > 0$  is a constant that depends on  $r$ . Hence, we obtain the following recurrence for  $T(n)$ :

$$T(n) \leq \begin{cases} c_1 r^{10} T(n/r^6) + c_3 n^{(\omega+1)/2} \log^{c_2} n & n \geq n_0, \\ c_4 & n < n_0, \end{cases} \quad (21)$$

where  $c_4 > 0$  and  $n_0 > 0$  are constants that depend on  $r$ . Note that the recursive subproblems may not be balanced but the total time spent in a round with subproblem size  $t \geq n_0$  is still at most  $c_3 t^{(\omega+1)/2} \log^{c_2} t$ , so the above recurrence holds.

Assuming  $r$  is chosen sufficiently large, the solution of the above recurrence is

$$T(n) \leq A n^\alpha \log^{c_2} n, \quad \text{where } \alpha = \max\{5/3 + \varepsilon, (\omega + 1)/2\} \quad (22)$$

for any constant  $\varepsilon > 0$ , where  $A$  is an appropriate constant that depends on  $r$ . Since the current best known value of  $\omega$  is roughly 2.37, we simply write the solution of the recurrence as  $T(n) = \tilde{O}(n^{(\omega+1)/2})$ . We remark that the first term in the definition of  $\alpha$  can be reduced to  $8/5 + \varepsilon$  by observing that points of  $\mathbf{R}^* \cup \mathbf{B}^*$  lie on the surface  $\gamma_i = z_i \alpha_i - \beta_i y_i$  and thus only  $O(r^4 \log r)$  cells of  $\Xi$  (resp.  $\Xi_\Delta$ ) contain a point of  $\mathbf{B}$  (resp.  $\mathbf{R}_\Delta$ ) (see e.g. [7, 1]), and thus we recursively solve  $O(r^8 \log^2 r)$  subproblems, each of size at most  $n/r^5$ , after the first round.

For general values of  $m$  and  $n$  with  $m \leq n$ , the batching technique implies that the total time spent in computing  $\Phi_C(\mathbf{R}, \mathbf{B})$  is  $(1 + n/m) \cdot \tilde{O}(m^{(\omega+1)/2}) = \tilde{O}(nm^{(\omega-1)/2})$ . For  $m > n$ , we flip the roles of  $\mathbf{R}$  and  $\mathbf{B}$ , and the run time is  $\tilde{O}(mn^{(\omega-1)/2})$ . Hence, we obtain the following:

**Lemma 4.** *For a biclique  $(R_k, B_k)$  in the biclique partition computed in the first stage, with  $|R_k| = m_k$  and  $|B_k| = n_k$ ,  $\Phi_C(R_k, B_k)$  can be computed in  $\tilde{O}(m_k n_k^{(\omega-1)/2} + n_k m_k^{(\omega-1)/2})$  time.*

To bound the total time spent in computing  $\Phi(\mathbf{E}_f, \mathbf{E}_g)$ , we sum the bound in Lemma 4 over all bicliques of  $\mathcal{B}$ . Using Hölder's inequality and property (iv) of the biclique partition  $\mathcal{B}$ , we obtain the following:

$$\begin{aligned} \sum_{k=1}^s m_k n_k^{(\omega-1)/2} &= \sum_{k=1}^s (m_k n_k)^{(\omega-1)/2} m_k^{(3-\omega)/2} \\ &\leq \left( \sum_{k=1}^s m_k n_k \right)^{(\omega-1)/2} \cdot \left( \sum_{k=1}^s m_k \right)^{(3-\omega)/2} \\ &= \tilde{O}(\kappa^{(\omega-1)/2} n^{(3-\omega)/2}), \end{aligned}$$

which is  $\tilde{O}(n)$  for  $\kappa = n$  and  $\tilde{O}(n^{(\omega+1)/2})$  for  $\kappa = n^2$ . Similarly, we can argue that  $n_k m_k^{(\omega-1)/2} = \tilde{O}(\kappa^{(\omega-1)/2} n^{(3-\omega)/2})$ . Adding the  $O(n \log n)$  time spent in evaluating (4) at the vertices of  $\mathbf{V}_f \cup \mathbf{V}_g$ , the total time spent in computing  $\|f - g\|_1$  is  $\tilde{O}(\kappa^{(\omega-1)/2} n^{(3-\omega)/2})$ . This proves Theorem 1.

## 4 Multipoint Evaluation of Bivariate Rational Functions

In this section, we show how computing the quantities  $\Phi_C(\mathbf{R}, \mathbf{B})$  associated to each  $m \times n$  biclique  $\mathbf{R} \times \mathbf{B}$  can be reduced to computing a double sum of  $m$  rational functions  $\frac{1}{q_i(x,y)}$  over  $n$  points  $p_j$ . This sum can then be evaluated using fast multipoint evaluation techniques for bivariate polynomials ([14]), based on fast modular composition ([18, Chapter 12, §12.2]). As in the previous section, the other terms appearing in  $\Phi(\mathbf{E}_f, \mathbf{E}_g)$  can be computed in an analogous manner. We start with a technical lemma on the fast multipoint evaluation of bivariate polynomials.

**Lemma 5.** *Given  $n$  polynomials  $f_i(x, y)$  of degree  $n$  and  $n^2$  points  $p_j \in \mathbb{R}^2$ , it is possible to evaluate all the  $f_i(p_j)$  in  $\tilde{O}(n^{\omega+1})$  arithmetic operations.*

*Proof.* The main idea is to reduce this question to a problem of fast multipoint evaluation of univariate polynomials.

First, we interpolate two polynomials  $g(x)$  and  $h(x)$  of degree  $n^2$ , such that, for each root  $r$  of  $h$ , the point  $(r, g(r))$  is a point  $p_i$ . More formally we require that

$$\{p_1, \dots, p_n\} = \{(x_i, g(x_i)) \mid h(x_i) = 0\}.$$

(For clarity of presentation, we assume that all the coordinates  $x_i$  are distinct; the general case can be handled by introducing a generic shear transformation of the  $xy$ -plane at no asymptotic cost; see [14, §6].) Polynomials  $g(x)$  and  $h(x)$  can be computed in  $\tilde{O}(n^2)$  operations using fast interpolation algorithms [18, Chapter 10.2].

Then, denoting by  $\tilde{f}_j(x)$  the univariate polynomial  $f_j(x, g(x)) \bmod h(x)$ , note that  $f_j(x_i, y_i) = \tilde{f}_j(x_i)$  for all  $i$ . Hence, if we know the coefficients of  $\tilde{f}_j$ , we can evaluate each  $\tilde{f}_j$  at all  $x_i$  in  $\tilde{O}(n^2)$  arithmetic operations, and all  $\tilde{f}_j$  at all  $x_i$  in  $\tilde{O}(n^3)$  operations.

Finally, computing all the  $\tilde{f}_j$  can be done in  $\tilde{O}(n^{\omega+1})$  arithmetic operations using Lemma 10(iii) of [14], which can be seen as a variation of the fast composition algorithm 12.3 in [18, Chapter 12, §12.2]. Thus, the final complexity is in  $\tilde{O}(n^{\max(3, \omega+1)}) = \tilde{O}(n^{\omega+1})$ .  $\square$

We can now deduce our corollaries for computing the sum of rational functions. First we start with the case where the numerators are all 1.

**Corollary 6.** *Given  $n$  constant-degree polynomials  $q_i(x, y)$  and  $n$  points  $p_j \in \mathbb{R}^2$ , the  $n$  sums  $\sum_i \frac{1}{q_i(p_j)}$  for all  $1 \leq j \leq n$  can be computed in  $\tilde{O}(n^{\frac{\omega+1}{2}})$  arithmetic operations.*

*Proof.* First, partitioning the set of indices  $i$  into  $O(\sqrt{n})$  subsets  $I_\ell$  of size  $O(\sqrt{n})$ , for each  $\ell$ , there exists a polynomial  $f_\ell(x, y)$  of degree  $O(\sqrt{n})$  such that  $\sum_{i \in I_\ell} \frac{1}{q_i(x, y)} = \frac{f_\ell(x, y)}{\prod_{i \in I_\ell} q_i(x, y)}$ . Each  $f_\ell$  can be computed in  $\tilde{O}(n)$  arithmetic operations, using fast bivariate multiplication [18, §8.4]. Then, using Lemma 5, we can evaluate all the  $O(\sqrt{n})$  polynomials  $f_\ell$  on all the  $n$  points  $p_j$  in  $\tilde{O}(n^{\frac{\omega+1}{2}})$  operations. With the same approach, we can expand the denominators and evaluate them on the points  $p_j$  in  $\tilde{O}(n^{\frac{\omega+1}{2}})$  operations.  $\square$

Then we generalize this result to sums of general rational functions.

**Corollary 7.** For positive integers  $k, n, i$ , with  $i \leq n$ , let  $f_i$  and  $g_i$  be polynomials of constant degree in  $k$  and two variables, respectively. Let  $A \subset \mathbb{R}^k$  contain at most  $n$  points. In the Real RAM model, the sum

$$S = \sum_{(a_1, \dots, a_k) \in A} \sum_{1 \leq i \leq n} \frac{f_i(a_1, \dots, a_k)}{g_i(a_1, a_2)}$$

can be computed in  $\tilde{O}(n^{\frac{\omega+1}{2}})$  arithmetic operations.

*Proof.* Given  $k$  symbolic variables  $X_1, \dots, X_k$ , all the polynomials  $f_i(X_1, \dots, X_k)$  can be written as a weighted sum of  $s$  monomials  $M_j$ , where  $M_j = X_1^{e_{j,1}} \cdots X_k^{e_{j,k}}$  with  $(e_{j,1}, \dots, e_{j,k}) \in \mathbb{N}^k$ . In particular, each polynomial  $f_i(X_1, \dots, X_k)$  can be expanded as

$$c_{i,1}M_1 + \cdots + c_{i,s}M_s.$$

According to Lemma 6, letting  $q_i = \frac{g_i}{c_{i,1}}$ , we can compute for all the points in  $(a_1, \dots, a_k) \in A$  the sums  $\sum_{i=1}^n \frac{1}{q_i(a_1, a_2)}$  in  $\tilde{O}(n^{\frac{\omega+1}{2}})$  operations. Since the number  $s$  of monomials is constant, we can compute with the same complexity for all  $1 \leq j \leq s$  and all points  $p = (a_1, \dots, a_k) \in A$  the sums :

$$S_{p,j} := \sum_{i=1}^n \frac{c_{i,j}}{g_i(a_1, a_2)}.$$

On the other hand, since  $s$  and the degrees of the  $f_i$  are constant, we can directly compute all the monomials  $M_j$  at all the points of  $A$  in  $O(n)$  operations. Using an additional number of operations linear in  $n$ , this allows us to compute, for all the points in  $p \in A$ ,

$$S_p := \sum_{i=1}^n \frac{f_i(a_1, \dots, a_k)}{g_i(a_1, a_2)} = \sum_{j=1}^s S_{p,j} M_j(p).$$

Finally, we get the desired result by summing the quantities  $S_p$  over all  $p \in A$ .  $\square$

This concludes the description of the multipoint evaluation tools needed in Section 3 to complete our algorithm.

## 5 Conclusion

A recent unpublished preprint improves the bound on multipoint evaluation for multivariate polynomials. It shows that we could evaluate  $n^{(\omega-1)}$  bivariate polynomials of degree  $n^2$  on  $n^2$  points in  $\tilde{O}(n^{\omega+1})$  operations [16, Theorem 3.8]. Compared to Lemma 5 derived from earlier work [14], this allows us to evaluate more polynomials with the same number of operations. In turn, that would improve Corollary 7, and the complexity to evaluate the sum of rational functions on a biclique of size  $n \times n$  would become  $\tilde{O}(n^{2-1/\omega})$  instead of  $O(n^{(\omega+1)/2})$ , by summing over batches of  $n^{2/\omega}$  points. With the current  $\omega$  value,  $2 - 1/\omega$  is approximately 1.58, which is below  $8/5$ . Thus the cost of computing the biclique decomposition would become dominant, and the total cost of computing  $\|f - g\|_1$  could be further improved to  $O(n^{8/5+\varepsilon})$ , for any constant  $\varepsilon > 0$ .

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