

## Interpolation by Piecewise-Linear Radial Basis Functions, II

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In the two-dimensional plane, a set of nodes  $x_1, x_2, \dots, x_n$  is given. It is desired to interpolate arbitrary data given at the nodes by a linear combination of the functions  $h_i(x) = \|x - x_i\|$ . Here the norm is the  $l_1$ -norm. For this purpose, one can employ the space  $\mathcal{PL}$  of all continuous piecewise-linear functions on the rectangular grid generated by the nodes. Interpolation at the nodes by this larger space is quite easy. By adding an appropriate  $\mathcal{PL}$ -function that vanishes on the nodes, we can obtain the linear combination of  $h_1, h_2, \dots, h_n$  that interpolates the data. This algorithm is much more efficient than the straightforward method of simply solving the linear system of equations  $\sum c_j h_j(x_i) = d_i$ . © 1991 Academic Press, Inc.

### 1. INTRODUCTION

Throughout the paper,  $\mathcal{N}$  denotes a set of  $n$  distinct points (*nodes*) in  $\mathbb{R}^2$  designated by  $x_1, x_2, \dots, x_n$ . The basic problem of two-dimensional interpolation addressed here is as follows. A "data-function"  $d: \mathcal{N} \rightarrow \mathbb{R}$  is given, and we seek a function  $f: \mathbb{R}^2 \rightarrow \mathbb{R}$  such that  $f|_{\mathcal{N}} = d$ ; i.e.,  $f(x_i) = d_i$  for  $i = 1, 2, \dots, n$ . Such a function  $f$  is said to *interpolate*  $d$ . Usually the search for  $f$  is restricted to a class of functions that (a) are easily computed and (b) have some prescribed smoothness.

We seek an interpolant in the linear space generated by the  $n$  functions  $h_j(x) = \|x - x_j\|$  ( $1 \leq j \leq n$ ), where the norm is chosen to be the  $l_1$ -norm.

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The existence of an interpolant  $f = \sum_{j=1}^n c_j h_j$  for arbitrary data depends upon the invertibility of the *interpolation matrix*  $A$ , whose elements are  $A_{ij} = h_j(x_i)$ . In [1] it was shown that a necessary and sufficient condition for the nonsingularity of  $A$  is that  $\mathcal{N}$  contain no closed rectilinear path.

Notation of [1] will be briefly reviewed here. If  $x$  is a point in  $\mathbb{R}^2$ , we display its coordinates by writing  $x = (s, t)$ . The nodes are  $x_i = (s_i, t_i)$ . Two coordinate projections are defined by  $Px = s$  and  $Qx = t$ . We set

$$P(\mathcal{N}) = \{\sigma_1, \sigma_2, \dots, \sigma_m\}, \quad \sigma_1 < \sigma_2 < \dots < \sigma_m$$

$$Q(\mathcal{N}) = \{\tau_1, \tau_2, \dots, \tau_k\}, \quad \tau_1 < \tau_2 < \dots < \tau_k.$$

The rectangular grid and the rectangular hull determined by  $\mathcal{N}$  are

$$RG = \{\sigma_1, \sigma_2, \dots, \sigma_m\} \times \{\tau_1, \tau_2, \dots, \tau_k\}$$

$$RH = \{(s, t): \sigma_1 \leq s \leq \sigma_m \text{ and } \tau_1 \leq t \leq \tau_k\}.$$

The horizontal lines  $t = \tau_j$  and the vertical lines  $s = \sigma_i$  divide the plane into  $(m+1)(k+1)$  rectangles, some of which are unbounded. The space  $\mathcal{P}\mathcal{L}$  consists of all continuous functions on  $\mathbb{R}^2$  that are linear on each of these rectangles. The space  $\mathcal{R}\mathcal{B}$  is the linear span of the set  $\{h_1, h_2, \dots, h_n\}$ . The functions  $h_i$  are defined by

$$h_i(x) = \|x - x_i\| = |s - s_i| + |t - t_i|.$$

A path is defined as an ordered finite set  $[y_1, y_2, \dots, y_q]$  in  $RG$  such that the line segments joining consecutive points have positive length and are alternately horizontal and vertical. A path is said to be closed if  $q$  is even, if  $y_q \neq y_1$ , and if the line segment joining  $y_1$  with  $y_2$  is perpendicular to the line segment joining  $y_1$  with  $y_q$ .

## 2. AN EQUIVALENCE RELATION

In this section, we begin an analysis which leads to a description of  $\mathcal{R}\mathcal{B}$  as a subspace of  $\mathcal{P}\mathcal{L}$ . The description is in the "dual" form, which is to say that  $\mathcal{R}\mathcal{B}$  will be exhibited as the intersection of hyperplanes.

**DEFINITION.** An equivalence relation is introduced in  $\mathcal{N}$  by declaring that two nodes are equivalent if there is a path in  $\mathcal{N}$  that connects them. We also declare each node equivalent to itself.

**EXAMPLE.** A set  $\mathcal{N}$  having two equivalence classes is shown in Fig. 2.1. The following elementary lemma is given without proof.

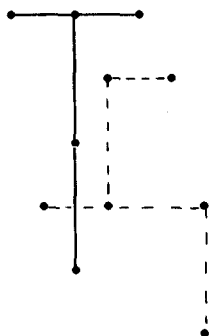


FIGURE 2.1

2.1. LEMMA. (a) If  $\mathcal{N}_0, \mathcal{N}_1, \dots, \mathcal{N}_r$  are the equivalence classes that compose  $\mathcal{N}$ , then the sets  $P(\mathcal{N}_i)$  are pairwise disjoint.

(b) These equivalences hold for  $1 \leq j \leq n, 0 \leq i \leq r$ :

$$s_j \in P(\mathcal{N}_i) \Leftrightarrow x_j \in \mathcal{N}_i \Leftrightarrow t_j \in Q(\mathcal{N}_i).$$

2.2. LEMMA. If  $\mathcal{N}$  contains no closed path and consists of just one equivalence class, then  $m + k = n + 1$ .

*Proof.* Since  $\mathcal{N}$  contains no closed path, we infer from 3.2 in [1] that some grid line generated by  $\mathcal{N}$  contains only one node. We can assume, without loss of generality, that the horizontal line through  $x_1$  contains no other node. We now construct a graph-theoretic tree having root  $x_1$ . At level 1 we place  $x_1$ . At level 2 we place all nodes (other than  $x_1$ ) that lie on the vertical line through  $x_1$ . At level 3 we place all nodes not on levels 2 or 1 which lie on horizontal lines through the nodes on level 2. This process is continued as long as possible. For a formal description, let  $L_i$  denote the set of nodes at level  $i$ . Then  $L_1 = \{x_1\}$ , and recursively we put

$$L_{i+1} = [\mathcal{N} \cap Q^{-1}(Q(L_i))] \setminus [L_1 \cup \dots \cup L_i] \quad \text{if } i \text{ is even}$$

$$L_{i+1} = [\mathcal{N} \cap P^{-1}(P(L_i))] \setminus [L_1 \cup \dots \cup L_i] \quad \text{if } i \text{ is odd.}$$

The connections in the tree are described as follows. A node  $v$  on level  $i + 1$  will be joined to a node  $u$  on level  $i$  if and only if the nodes  $u$  and  $v$  lie on the same horizontal line (when  $i$  is even) or on the same vertical line (when  $i$  is odd). Every path in  $\mathcal{N}$  that starts at  $x_1$  can be traced through the successive levels of the tree. Since every node is connected to  $x_1$  by a path, every node is in the tree.

We shall now prove that each point of  $L_i$  ( $i > 1$ ) accounts for one new grid line. Suppose, on the contrary, that a node  $y_0$  in  $L_i$  does not generate

a new line. Say  $i$  is odd, so that the new grid lines generated by points of  $L_i$  are vertical. Then there exists a node  $z_0 \in L_j$ , with  $j \leq i$ ,  $z_0 \neq y_0$ , and  $P(z_0) = P(y_0)$ . By tracing backwards through the tree from  $z_0$  to  $y_0$  we eventually arrive at a first common node (which may be  $x_1$ ). This process generates two paths

$$[z_0, z_1, \dots, z_p] \quad \text{and} \quad [y_0, y_1, \dots, y_q]$$

in which  $z_p$  and  $y_q$  are the same point in the tree. Since  $z_p$  is the first common node,  $z_{p-1} \neq y_{q-1}$ . Then

$$[z_0, z_1, \dots, z_{p-1}, y_{q-1}, y_{q-2}, \dots, y_0]$$

is a path. An application of 3.6 in [1] shows that this path contains a closed subpath, contrary to hypothesis.

Since each node occurs exactly once in the tree and each node generates one grid line (except for  $x_1$  which generates two), the number of grid lines is  $n + 1$ , the number of horizontal lines is  $k$ , and the number of vertical lines is  $m$ , and hence  $n + 1 = m + k$  (see Fig. 2.2). ■

2.3. THEOREM. *Let  $\mathcal{N}$  be a node set having  $n$  points and containing no closed path. If  $r + 1$  denotes the number of equivalence classes in  $\mathcal{N}$ , then  $\dim \mathcal{P}\mathcal{L} = n + r + 4$ .*

*Proof.* By definition,  $m = \#P(\mathcal{N})$  and  $k = \#Q(\mathcal{N})$ . By applying 2.2 to each equivalence class  $\mathcal{N}_i$  we obtain

$$\#P(\mathcal{N}_i) + \#Q(\mathcal{N}_i) = \#\mathcal{N}_i + 1.$$

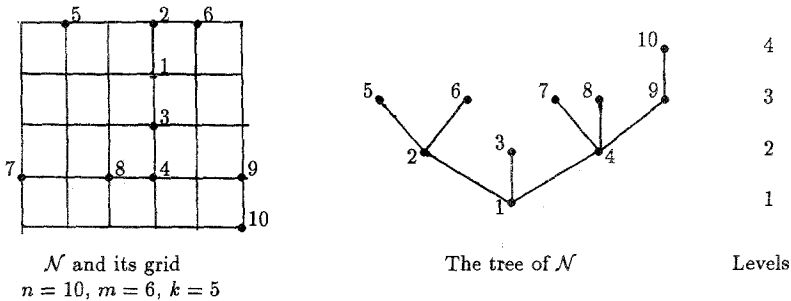


FIG. 2.2. A node set and its tree.

Since the set  $P(\mathcal{N}_i)$  are pairwise disjoint and since the same is true of the sets  $Q(\mathcal{N}_i)$ , we have, by 2.1 in [1],

$$\begin{aligned} \dim \mathcal{P}\mathcal{L} &= m + k + 3 = \sum_{i=0}^r [\#P(\mathcal{N}_i) + \#Q(\mathcal{N}_i)] + 3 \\ &= \sum_{i=0}^r [\#\mathcal{N}_i + 1] + 3 = n + r + 4. \quad \blacksquare \end{aligned}$$

### 3. ANNIHILATING FUNCTIONALS FOR $\mathcal{RB}$

The space  $\mathcal{RB}$  generated by the functions  $x \mapsto \|x - x_j\|$  is a subspace of  $\mathcal{P}\mathcal{L}$ . In this section we describe  $\mathcal{RB}$  in the dual manner—that is, as a family of functions in  $\mathcal{P}\mathcal{L}$  that satisfy a set of homogeneous linear equations. To this end, we shall define a set of functionals  $\Delta_0, \dots, \Delta_{r+4}$  which annihilate  $\mathcal{RB}$ .

All the notation previously defined is retained here, and in addition we set

$$\begin{aligned} \lambda^{-1} &= (\sigma_m - \sigma_1) + (\tau_k - \tau_1) \\ \sigma_0 &= \sigma_1 - \lambda^{-1}, & \sigma_{m+1} &= \sigma_m + \lambda^{-1} \\ \tau_0 &= \tau_1 - \lambda^{-1}, & \tau_{k+1} &= \tau_k + \lambda^{-1}. \end{aligned}$$

Two definitions are given next, along with some elementary consequences without proofs.

3.1. DEFINITION. For each  $\sigma \in P(\mathcal{N})$ , we define a linear functional  $\psi_\sigma$  which can act on any univariate piecewise linear function:

$$\psi_\sigma(u) = \lim_{s \downarrow \sigma} u'(s) - \lim_{s \uparrow \sigma} u'(s).$$

3.2. LEMMA. For the function  $u(s) = |s - \alpha|$ , we have  $\psi_\sigma(u) = 2$  if  $\alpha = \sigma$  and  $\psi_\sigma(u) = 0$  if  $\alpha \neq \sigma$ .

3.3. DEFINITION. For  $0 \leq i \leq r$  we define

$$\Psi_i = \sum \{\psi_\sigma : \sigma \in P(\mathcal{N}_i)\}.$$

3.4. LEMMA. Let  $u(s) = |s - \alpha|$ . Then  $\Psi_i(u) = 2$  if  $\alpha \in P(\mathcal{N}_i)$  and  $\Psi_i(u) = 0$  if  $\alpha \notin P(\mathcal{N}_i)$ .

In the same manner, we define functionals  $\theta_\tau$  and  $\Theta_i$  acting on functions of  $t$ . Then we define  $\Delta_i$  on  $\mathcal{P}\mathcal{L}(\mathcal{N})$  as follows. Given  $f \in \mathcal{P}\mathcal{L}$ , write

$$f(s, t) = u(s) + v(t)$$

with  $u \in \mathcal{P}\mathcal{L}(P(\mathcal{N}))$  and  $v \in \mathcal{P}\mathcal{L}(Q(\mathcal{N}))$ . (This expression is not unique.) Then define

$$\Delta_i(f) = \Psi_i(u) - \Theta_i(v) \quad (0 \leq i \leq r).$$

The definition is proper, for if another expression for  $f$  is chosen it must be of the form

$$f(s, t) = [u(s) + c] + [v(s) - c]$$

for some constant  $c$ . But  $\psi_\sigma(u + c) = \psi_\sigma(u)$ , since  $\psi_\sigma$  measures the jump in the derivative at  $\sigma$ . Hence  $\Psi_i(u + c) = \Psi_i(u)$  and similarly  $\Theta_i(v - c) = \Theta_i(v)$ .

3.5. LEMMA. Each functional  $\Delta_i$ , for  $0 \leq i \leq r$ , annihilates  $\mathcal{R}\mathcal{B}$ .

*Proof.* It suffices to prove that  $\Delta_i(h_j) = 0$ , where

$$h_j(x) = \|x - x_j\| = |s - s_j| + |t - t_j| = u(s) + v(s) \quad (1 \leq j \leq n).$$

If  $x_j \in \mathcal{N}_\alpha$  then by 2.1,  $s_j$  belongs only to  $P(\mathcal{N}_\alpha)$  and  $t_j$  belongs only to  $Q(\mathcal{N}_\alpha)$ . Hence either  $\Psi_i(u) = \Theta_i(v) = 2$  or  $\Psi_i(u) = \Theta_i(v) = 0$ . Consequently  $\Delta_i(h_j) = 0$ . ■

We now define four additional functionals  $\Delta_i$  (for  $r + 1 \leq i \leq r + 4$ ) which annihilate  $\mathcal{R}\mathcal{B}$ . Points  $y_i$  and  $z_i$  are defined as in Fig. 3.1 (which is *not* drawn to scale). The points  $z_1, \dots, z_4$  are at the corners of the rectangular

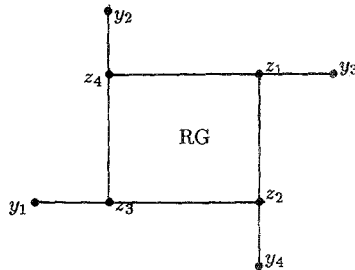


FIGURE 3.1

grid. Hence  $z_1 = (\sigma_m, \tau_k)$ ,  $z_2 = (\sigma_m, \tau_1)$ ,  $z_3 = (\sigma_1, \tau_1)$ , and  $z_4 = (\sigma_1, \tau_k)$ . The points  $y_1, \dots, y_4$  are situated as shown, and satisfy

$$\|y_1 - z_3\|_1 = \|y_2 - z_4\|_1 = \|y_3 - z_1\|_1 = \|y_4 - z_2\|_1 = \|z_2 - z_4\|_1 = \lambda^{-1}.$$

Now we put, for  $1 \leq i \leq 4$ ,

$$\Delta_{r+i} = \hat{y}_i + \hat{z}_i - (\hat{z}_1 + \hat{z}_2 + \hat{z}_3 + \hat{z}_4).$$

The circumflex signifies a point-evaluation functional.

3.6. LEMMA. *The four functionals  $\Delta_{r+1}, \dots, \Delta_{r+4}$  annihilate  $\mathcal{RB}$ .*

*Proof.* We wish to show that  $\Delta_{r+i}(h_j) = 0$  for  $j = 1, \dots, n$ . Fixing  $j$ , we observe that

$$\sum_{i=1}^4 h_j(z_i) = \sum_{i=1}^4 \|z_i - x_j\|_1 = \text{perimeter of } RG.$$

For a fixed  $i$ , let  $z_\alpha$  be the vertex opposite  $z_i$ . Then

$$\begin{aligned} h_j(y_i) + h_j(z_i) &= \|y_i - x_j\|_1 + \|z_i - x_j\|_1 \\ &= \|y_i - z_\alpha\|_1 + \|z_\alpha - x_j\|_1 + \|x_j - z_i\|_1 \\ &= \|z_2 - z_4\|_1 + \|z_2 - z_4\|_1 \\ &= \text{perimeter of } RG. \end{aligned}$$

By the definition of  $\Delta_{r+i}$ , we have  $\Delta_{r+i}(h_j) = 0$ . ■

The proof that  $\{\Delta_0, \Delta_1, \dots, \Delta_{r+4}\}$  spans  $\mathcal{RB}^\perp$  is deferred to Section 5.

#### 4. THE SUBSPACE $\mathcal{M}$

4.1. DEFINITIONS. The subspace  $\mathcal{M}$  is defined by

$$\mathcal{M} = \{f \in \mathcal{PL}(\mathcal{N}) : f|_{\mathcal{N}} = 0\}.$$

Further definitions follow. Functions  $u_0, u_1, \dots, u_r$  are defined in  $\mathcal{PL}(\mathbb{R})$  by specifying their knots to be  $\sigma_1, \dots, \sigma_m$  and specifying their values to be

$$u_i(\sigma_j) = \begin{cases} 1, & \sigma_j \in \mathcal{P}(\mathcal{N}_i) \\ 0, & \text{otherwise} \end{cases} \quad (0 \leq j \leq m+1).$$

A typical function  $u_i$  is graphed in Fig. 4.1. Notice that  $u_i(\sigma_0) = u_i(\sigma_{m+1}) = 0$  for all  $i$ .

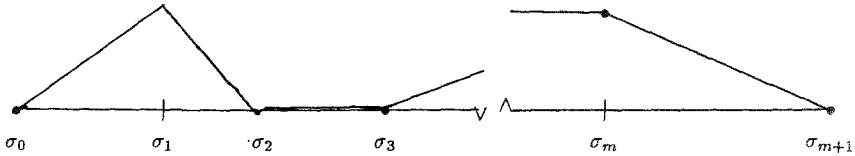


FIGURE 4.1

Functions  $v_0, v_1, \dots, v_r$  are defined to be piecewise linear with knots at  $\tau_1, \dots, \tau_k$ , and having these values

$$v_i(\tau_j) = \begin{cases} 1 & \text{if } \tau_j \in Q(\mathcal{N}_i) \\ 0 & \text{otherwise} \end{cases} \quad (0 \leq j \leq k + 1).$$

Then we define  $g_i(s, t) = u_i(s) - v_i(t)$  for  $0 \leq i \leq r$ .

4.2. LEMMA. *The functions  $g_0, \dots, g_r$  belong to  $\mathcal{M}$ .*

*Proof.* Let  $x_j$  be any node. Let  $\mathcal{N}_\alpha$  be the equivalence class containing  $x_j$ . Then  $s_j \in P(\mathcal{N}_\alpha)$  and  $t_j \in Q(\mathcal{N}_\alpha)$ , by 2.1. We conclude that  $g_i(x_j) = 0$ . ■

We now define four additional  $\mathcal{P}\mathcal{L}$  functions  $g_i$  for  $r + 1 \leq i \leq r + 4$ . Each of these vanishes on  $RG$ . At the special points  $y_j$  we assign these values:

$$g_{r+i}(y_j) = \delta_{ij}, \quad 1 \leq i, j \leq 4.$$

It is clear that each of these functions belongs to  $\mathcal{M}$  and the set  $\{g_{r+1}, \dots, g_{r+4}\}$  is linearly independent. Observe also that  $\Delta_{r+i}(g_{r+j}) = \delta_{ij}$ , by the definition of  $\Delta_{r+i}$  in Section 3.

4.3. LEMMA. *The four functions  $g_{r+1}, \dots, g_{r+4}$  form a basis for the subspace of  $\mathcal{P}\mathcal{L}$  consisting of functions which vanish on the rectangular grid.*

*Proof.* Let  $f$  be a  $\mathcal{P}\mathcal{L}$ -function such that  $f|_{RG} = 0$ . Put  $c_i = f(y_i)$ . We assert that  $f = \sum_1^4 c_i g_{r+i}$ . For points in  $RG$  this is true since each  $g_{r+i}$  vanishes on  $RG$ . For points outside  $RG$ , we use the fact that the values of  $f$  at three vertices of a rectangle determine its value at the fourth vertex. With this, we see that  $f$  is completely determined by the four values  $f(y_1), \dots, f(y_4)$ . For example, the values  $f(y_1), f(\sigma_1, \tau_1) = f(\sigma_1, \tau_2) = 0$  determine  $f$  in the strip where  $\tau_1 \leq t \leq \tau_2$  and  $s \leq \sigma_1$ . ■

4.4. LEMMA. *The function  $\bar{g} = g_0 + \dots + g_r$  vanishes on the rectangular grid.*



*Proof.* Observe that  $\bar{g} \in \mathcal{P}\mathcal{L}$ . By 2.1 in [1], it suffices to prove that  $\bar{g}$  vanishes at the points

$$(\sigma_1, \tau_j), \quad (\sigma_\mu, \tau_1) \quad (1 \leq j \leq k, 1 \leq \mu \leq m).$$

We have

$$\bar{g}(\sigma_1, \tau_j) = \sum_{i=0}^r u_i(\sigma_1) - \sum_{i=0}^r v_i(\tau_j) = 1 - 1 = 0$$

because  $\sigma_1$  belongs to exactly one set  $P(\mathcal{N}_\alpha)$  and  $u_\alpha(\sigma_1) = 1$ , while all other  $u_i(\sigma_1) = 0$ . Similarly,  $\sum_{i=0}^r v_i(\tau_j) = 1$ . Analogous arguments are used at the other points. ■

4.5. LEMMA. *The set  $\{g_1, g_2, \dots, g_{r+4}\}$  is linearly independent.*

*Proof.* Suppose that  $\sum_{i=1}^r c_i g_i - \sum_{i=1}^4 a_i g_{r+i} = 0$ . It follows that

$$\sum_{i=1}^r c_i g_i |RG = \sum_{i=1}^4 a_i g_{r+i} |RG = 0.$$

Select a point  $(s, t)$  with  $s \in P(N_j)$ ,  $j \neq 0$ ,  $t \in Q(\mathcal{N}_0)$ . Then

$$0 = \sum_{i=1}^r c_i g_i(s, t) = \sum_{i=1}^r c_i [u_i(s) - v_i(t)] = c_j.$$

Next evaluate  $\sum_{i=1}^4 a_i g_{r+i}$  at  $y_j$  to see that  $a_j = 0$ . ■

4.6. LEMMA. *Let  $f$  be a  $\mathcal{P}\mathcal{L}$ -function which is constant on some equivalence class,  $\mathcal{N}_i$ . Let  $f(s, t) = u(s) + v(t)$ . Then  $u$  is constant on  $P(\mathcal{N}_i)$  and  $v$  is constant on  $Q(\mathcal{N}_i)$ .*

*Proof.* Let  $s$  and  $s'$  be any two points of  $P(\mathcal{N}_i)$ . Then there exist points  $t$  and  $t'$  such that  $(s, t) \in \mathcal{N}_i$  and  $(s', t') \in \mathcal{N}_i$ . By the definition of an equivalence class, the points  $(s, t)$  and  $(s', t')$  are connected by an open path whose vertices lie in  $\mathcal{N}_i$ . On any horizontal segment of this path, say from  $(\sigma, \tau)$  to  $(\sigma', \tau)$ , we have (since  $f$  is constant on  $\mathcal{N}_i$ )

$$u(\sigma) + v(\tau) = u(\sigma') + v(\tau)$$

whence  $u(\sigma) = u(\sigma')$ . On any vertical segment also the value of  $u$  does not change. Thus as the path is traversed, the value of  $u$  does not change, and  $u(s) = u(s')$ . Similarly,  $v$  is constant on  $Q(\mathcal{N}_i)$ . ■

4.7. THEOREM. Suppose that the node set  $\mathcal{N}$  has exactly  $n$  points, contains no closed path, and has  $r + 1$  equivalence classes. Then

- (i)  $\mathcal{PL} = \mathcal{RB} \oplus \mathcal{M}$
- (ii)  $\{g_1, g_2, \dots, g_{r+4}\}$  is a basis for  $\mathcal{M}$
- (iii)  $\sum_{i=0}^{r+1} g_i - g_{r+2} + g_{r+3} - g_{r+4} = 0$ .

*Proof.* Obviously  $\mathcal{RB} + \mathcal{M} \subset \mathcal{PL}$ . If  $f \in \mathcal{PL}$  then by 7.8 in [1] there is a unique  $h \in \mathcal{RB}$  that interpolates  $f$  on  $\mathcal{N}$ . Hence  $f - h \in \mathcal{M}$  and  $f \in \mathcal{RB} + \mathcal{M}$ . Thus  $\mathcal{PL} = \mathcal{RB} + \mathcal{M}$ . That  $\mathcal{RB} \cap \mathcal{M} = 0$  follows from 7.8 in [1], since the only  $\mathcal{RB}$  interpolant for zero data on  $\mathcal{N}$  is the 0-element of  $\mathcal{RB}$ .

From 2.3,  $\dim \mathcal{PL} = n + r + 4$ . Since  $\dim \mathcal{RB} = n$ , we have  $\dim \mathcal{M} = r + 4$ . By 4.5,  $\{g_1, \dots, g_{r+4}\}$  is linearly independent and therefore is a basis for  $\mathcal{M}$ .

Now let  $\bar{g} = \sum_0^r g_i$ . By 4.4,  $\bar{g}$  is a  $\mathcal{PL}$ -function which vanishes on the rectangular grid. By 4.3, there exist coefficients  $\alpha_{r+i}$  such that  $\bar{g} = \sum_1^4 \alpha_{r+i} g_{r+i}$ . By evaluating at the four points  $y_1, \dots, y_4$  we find that  $\alpha_{r+i} = (-1)^i$ . ■

### 5. A DUAL ALGORITHM FOR $\mathcal{RB}$ -INTERPOLATION

The direct method of computing an  $\mathcal{RB}$ -interpolant to a data function simply solves the interpolation equations

$$\sum_{j=1}^n a_j \|x_i - x_j\|_1 = d_i \quad (1 \leq i \leq n).$$

An alternative method proceeds by first solving the interpolation problem with a function  $f$  in  $\mathcal{PL}$ . One can use the method of Section 4 in [1] to do this. By 4.7,  $f$  has a unique representation

$$f = h + g, \quad h \in \mathcal{RB}, g \in \mathcal{M}.$$

Also by 4.7,  $g$  is expressible in terms of the functions  $g_i$ , say  $g = \sum_{j=1}^{r+4} a_j g_j$ . The coefficients  $a_j$  can be determined from the condition  $f - g \in \mathcal{RB}$ , which is equivalent to

$$\Delta_i(f - g) = 0, \quad 1 \leq i \leq r + 4$$

by 5.10 infra. These equations lead to the system

$$\sum_{j=1}^{r+4} \Delta_i(g_j) a_j = \Delta_i(f) \quad (1 \leq i \leq r + 4).$$

The invertibility of the matrix  $(\Delta_i(g_j))$  follows from 7.8 in [1].

In order to carry out this dual algorithm, it will be necessary to evaluate the elements  $\Delta_i(g_j)$  in the coefficient matrix. In this section these elements are computed.

Recall the definitions of  $\Delta_i$  and  $g_j$  given in Sections 3 and 4. In addition, define functionals  $\psi_\sigma^+$  and  $\psi_\sigma^-$  for any  $\sigma \in \mathbb{R}$  by the equations

$$\psi_\sigma^+(u) = \lim_{s \downarrow \sigma} u'(s)$$

$$\psi_\sigma^-(u) = \lim_{s \uparrow \sigma} u'(s).$$

5.1. LEMMA. *If  $0 \leq i \leq r$ , then the value of  $\Psi_i(u_i)$  is the sum of all terms  $(\sigma_{\mu-1} - \sigma_\mu)^{-1}$  for which either*

- (i)  $\sigma_\mu \in P(\mathcal{N}_i)$  and  $\sigma_{\mu-1} \notin P(\mathcal{N}_i)$  or
- (ii)  $\sigma_\mu \notin P(\mathcal{N}_i)$  and  $\sigma_{\mu-1} \in P(\mathcal{N}_i)$ .

*Proof.* If  $\sigma_\mu \in P(\mathcal{N}_i)$  then

$$\begin{aligned} \psi_{\sigma_\mu}^+(u_i) &= [u_i(\sigma_{\mu+1}) - u_i(\sigma_\mu)](\sigma_{\mu+1} - \sigma_\mu)^{-1} \\ &= \begin{cases} (\sigma_\mu - \sigma_{\mu+1})^{-1} & \text{if } \sigma_{\mu+1} \notin P(\mathcal{N}_i) \\ 0 & \text{if } \sigma_{\mu+1} \in P(\mathcal{N}_i). \end{cases} \end{aligned}$$

Similarly,

$$\psi_{\sigma_\mu}^-(u_i) = \begin{cases} (\sigma_\mu - \sigma_{\mu-1})^{-1} & \text{if } \sigma_{\mu-1} \notin P(\mathcal{N}_i) \\ 0 & \text{if } \sigma_{\mu-1} \in P(\mathcal{N}_i). \end{cases}$$

It follows that

$$\begin{aligned} \Psi_i(u_i) &= \sum \{ \psi_\sigma(u_i) : \sigma \in P(\mathcal{N}_i) \} \\ &= \sum \{ \psi_\sigma^+(u_i) : \sigma \in P(\mathcal{N}_i) \} - \sum \{ \psi_\sigma^-(u_i) : \sigma \in P(\mathcal{N}_i) \} \\ &= \sum \{ (\sigma_\mu - \sigma_{\mu+1})^{-1} : \sigma_\mu \in P(\mathcal{N}_i) \text{ and } \sigma_{\mu+1} \notin P(\mathcal{N}_i) \} \\ &\quad - \sum \{ (\sigma_\mu - \sigma_{\mu-1})^{-1} : \sigma_\mu \in P(\mathcal{N}_i) \text{ and } \sigma_{\mu-1} \notin P(\mathcal{N}_i) \} \\ &= \sum \{ (\sigma_{\mu-1} - \sigma_\mu)^{-1} : \sigma_{\mu-1} \in P(\mathcal{N}_i) \text{ and } \sigma_\mu \notin P(\mathcal{N}_i) \} \\ &\quad + \sum \{ (\sigma_{\mu-1} - \sigma_\mu)^{-1} : \sigma_\mu \in P(\mathcal{N}_i) \text{ and } \sigma_{\mu-1} \notin P(\mathcal{N}_i) \}. \quad \blacksquare \end{aligned}$$

5.2. LEMMA. *If  $0 \leq i, j \leq r$  and  $i \neq j$ , then  $\Psi_i(u_j)$  is the sum of all terms  $(\sigma_\mu - \sigma_{\mu-1})^{-1}$  where either*

- (i)  $\sigma_\mu \in P(\mathcal{N}_i)$  and  $\sigma_{\mu-1} \in P(\mathcal{N}_j)$  or
- (ii)  $\sigma_\mu \in P(\mathcal{N}_j)$  and  $\sigma_{\mu-1} \in P(\mathcal{N}_i)$ .

*Proof.* If  $\sigma_\mu \in P(\mathcal{N}_i)$  then

$$\psi_{\sigma_\mu}^+(u_j) = u_j(\sigma_{\mu+1})(\sigma_{\mu+1} - \sigma_\mu)^{-1}.$$

This is  $(\sigma_{\mu+1} - \sigma_\mu)^{-1}$  if  $\sigma_{\mu+1} \in P(\mathcal{N}_j)$  and is 0 otherwise. Hence as in the preceding proof

$$\begin{aligned} & \sum \{ \psi_\sigma^+(u_j) : \sigma \in P(\mathcal{N}_i) \} \\ &= \sum \{ (\sigma_{\mu+1} - \sigma_\mu)^{-1} : \sigma_\mu \in P(\mathcal{N}_i) \text{ and } \sigma_{\mu+1} \in P(\mathcal{N}_j) \} \\ &= \sum \{ (\sigma_\mu - \sigma_{\mu-1})^{-1} : \sigma_{\mu-1} \in P(\mathcal{N}_i) \text{ and } \sigma_\mu \in P(\mathcal{N}_j) \}. \end{aligned}$$

The calculation of  $\sum \{ \psi_\sigma^-(u_j) : \sigma \in P(\mathcal{N}_i) \}$  is similar. ■

All of what has been proved for the matrix with elements  $\Psi_i(u_j)$  can be proved for  $\Theta_i(v_j)$ , mutatis mutandis. Then for  $0 \leq i, j \leq r$ ,

$$\Delta_i(g_j) = \Delta_i(u_j - v_j) = \Psi_i(u_j) - \Theta_i(-v_j) = \Psi_i(u_j) + \Theta_i(v_j).$$

In what follows  $\Delta_i(g_j)$  will be computed in the remaining cases,  $r < i \leq r+4$  and  $r < j \leq r+4$ . It is necessary to single out the equivalence classes which contain nodes on the boundary of  $RH$ . To do so, we define integers  $\varepsilon_v$  for  $0 \leq v \leq 4$  by the equations

$$\varepsilon_0 = \varepsilon_4, \quad \sigma_1 \in P(\mathcal{N}_{\varepsilon_4}), \quad \sigma_m \in P(\mathcal{N}_{\varepsilon_2}), \quad \tau_1 \in Q(\mathcal{N}_{\varepsilon_3}), \quad \tau_k \in Q(\mathcal{N}_{\varepsilon_1}).$$

The indices  $\varepsilon_1, \varepsilon_2, \varepsilon_3, \varepsilon_4$  are not necessarily distinct.

5.3. LEMMA. *The following formula is valid*

$$\Delta_i(g_{r+v}) = \lambda(-1)^{v-1} \delta(i, \varepsilon_{v-1}) \quad (0 \leq i \leq r, 1 \leq v \leq 4).$$

*In this formula,  $\lambda = \|z_2 - z_4\|^{-1}$  and  $\delta$  is the Kronecker symbol.*

*Proof.* Recall that  $g_{r+v}$  is a  $\mathcal{P}\mathcal{L}$ -function which vanishes on  $RG$  and takes these values:

$$g_{r+v}(y_i) = \delta_{iv} \quad (1 \leq i, v \leq 4).$$

This function can be expressed in the form

$$g_{r+v}(s, t) = u_{r+v}(s) - v_{r+v}(t) \quad (1 \leq v \leq 4).$$

The functions on the right in this equation take zero values on  $\sigma_j$  and  $\tau_j$  except in these cases:

$$u_{r+1}(\sigma_0) = 1, \quad v_{r+2}(\tau_{k+1}) = -1, \quad u_{r+3}(\sigma_{m+1}) = 1, \quad v_{r+4}(\tau_0) = -1.$$

Consequently we have

$$\begin{aligned} \psi_{\sigma_1}(u_{r+1}) &= (\sigma_1 - \sigma_0)^{-1} = \lambda, & \psi_{\sigma_m}(u_{r+3}) &= \lambda, \\ \theta_{\tau_1}(v_{r+4}) &= -\lambda, & \theta_{\tau_k}(v_{r+2}) &= -\lambda. \end{aligned}$$

In all other cases,  $\psi_{\sigma_i}(u_{r+v}) = \theta_{\tau_j}(v_{r+v}) = 0$ . Hence

$$\Psi_i(u_{r+1}) = \begin{cases} \lambda & \text{if } \sigma_1 \in P(\mathcal{N}_i) \\ 0 & \text{if } \sigma_1 \notin P(\mathcal{N}_i). \end{cases}$$

Likewise,  $\Theta_i(v_{r+1}) = 0$ , and so

$$\Delta_i(g_{r+1}) = \Psi_i(u_{r+1}) + \Theta_i(v_{r+1}) = \lambda \delta(i, \varepsilon_4).$$

In a similar fashion we find that

$$\begin{aligned} \Delta_i(g_{r+2}) &= \Theta_i(v_{r+2}) = -\lambda \delta(i, \varepsilon_1) \\ \Delta_i(g_{r+3}) &= \Psi_i(u_{r+3}) = \lambda \delta(i, \varepsilon_2) \\ \Delta_i(g_{r+4}) &= \Theta_i(v_{r+4}) = -\lambda \delta(i, \varepsilon_3). \end{aligned}$$

Hence the formula in the statement of the lemma is valid, with the interpretation that  $\varepsilon_0 = \varepsilon_4$ . ■

5.4. LEMMA. For  $1 \leq v \leq 4$  and  $0 \leq j \leq r$  we have

$$C_{vj} = \Delta_{r+v}(g_j) = -\bar{\mu}_j + (-1)^v \delta(j, \varepsilon_{v-1})$$

in which  $\bar{\mu}_j = \sum_{v=1}^4 (-1)^v \delta(j, \varepsilon_v)$ .

*Proof.* One can verify these formulae:

$$\begin{aligned} g_j(y_v) &= (-1)^v \delta(j, \varepsilon_{v+2}) \\ g_j(z_v) &= (-1)^v \delta(j, \varepsilon_v) + (-1)^{v+1} \delta(j, \varepsilon_{v+1}) \\ g_j(z_1) + g_j(z_2) + g_j(z_3) + g_j(z_4) &= 2\bar{\mu}_j. \end{aligned}$$

For example, when  $v = 1$ , the first formula is proved by observing that

$$g_j(y_1) \neq 0 \Rightarrow \tau_1 \in Q(\mathcal{N}_j) \Rightarrow j = \varepsilon_3 \Rightarrow g_j(y_1) = -v_j(\tau_1) = -1.$$

When  $v = 3$ , the second formula is established as follows. Observe that  $g_j(z_3) \neq 0$  only if  $j = \varepsilon_3$  or  $j = \varepsilon_4$ . If  $j = \varepsilon_3 \neq \varepsilon_4$ , then  $g_j(z_3) = u_j(\sigma_1) - v_j(\tau_1) = 0 - 1 = -1$ . If  $j = \varepsilon_4 \neq \varepsilon_3$ , then  $g_j(z_3) = u_j(\sigma_1) = 1$ . If  $j = \varepsilon_3 = \varepsilon_4$  then  $g_j(z_3) = 1 - 1 = 0$ . With these formulae established, one verifies easily the assertion of the lemma. ■

5.5. LEMMA. If  $\sum_{i=0}^{r+4} b_i A_i = 0$  then

$$b_{r+1} + \lambda b_{e_4} = b_{r+2} - \lambda b_{e_1} = b_{r+3} + \lambda b_{e_2} = b_{r+4} - \lambda b_{e_3} = 0.$$

*Proof.* We prove just one of these, viz.  $b_{r+2} - \lambda b_{e_1} = 0$ . The others are similar. Construct a function  $v$  having the appearance shown in Fig. 5.1. Then put  $f(s, t) = v(t)$ . We have

$$A_{r+v}(f) = \begin{cases} 1 & \text{if } v = 2 \\ 0 & \text{otherwise } (1 \leq v \leq 4) \end{cases}$$

$$A_i(f) = -\Theta_i(v) = \begin{cases} -\lambda & \text{if } i = e_1 \\ 0 & \text{otherwise } (0 \leq i \leq r). \end{cases}$$

Consequently

$$0 = \sum_{i=0}^{r+4} b_i A_i(f) = -\lambda b_{e_1} + b_{r+2}. \quad \blacksquare$$

5.6. LEMMA. If  $\sum_{i=0}^{r+4} b_i A_i = 0$ , then

$$b_{r+1} + b_{r+2} + b_{r+3} + b_{r+4} = (b_{e_2} - b_{e_4}) / (\sigma_m - \sigma_1).$$

*Proof.* Consider the function  $u \in \mathcal{P}\mathcal{L}(\mathbb{R})$  whose graph is shown in Fig. 5.2. Put  $f(s, t) = u(s)$ . Then

$$f(z_3) = f(z_4) = f(y_1) = f(y_2) = 1$$

$$f(z_2) = f(z_1) = f(y_3) = f(y_4) = 0.$$

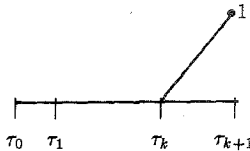


FIGURE 5.1

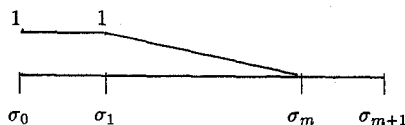


FIGURE 5.2

Consequently  $\Delta_{r+v}(f) = -1$  for  $1 \leq v \leq 4$ . For  $0 \leq i \leq r$  we have  $\Delta_i(f) = 0$  in all cases with two exceptions, namely,  $\Delta_{\varepsilon_2}(f) = -\Delta_{\varepsilon_4}(f) = (\sigma_m - \sigma_1)^{-1}$ . Thus

$$0 = \sum_{i=0}^{r+4} b_i \Delta_i(f) = - \sum_{v=1}^4 b_{r+v} + (b_{\varepsilon_2} - b_{\varepsilon_4})(\sigma_m - \sigma_1)^{-1}. \quad \blacksquare$$

5.7. LEMMA. If  $r+1 < n$  and if  $\sum_0^{r+4} b_i \Delta_i = 0$  then  $b_0 = b_1 = \dots = b_r$ .

*Proof.* The hypothesis states that the number of equivalence classes is less than  $n$ . Consequently some equivalence class contains at least two elements. By a renumbering of the equivalence classes we can assume that  $\#\mathcal{N}_0 \geq 2$ . Either  $\#P(\mathcal{N}_0) \geq 2$  or  $\#Q(\mathcal{N}_0) \geq 2$ , and we assume the former. Let  $a, b \in P(\mathcal{N}_0)$ , with  $a < b$ . Select any  $j \in \{1, \dots, r\}$ . We shall prove that  $b_j = b_0$ . Select  $c \in P(\mathcal{N}_j)$ .

Case 1. Assume  $c < a$ . Define  $f$  by

$$f(s, t) = u(s) = \begin{cases} 0, & s \leq c \\ (a-c)^{-1}(s-c), & c < s < a \\ -(b-a)^{-1}(s-b), & a < s < b \\ 0, & s \geq b. \end{cases}$$

Then  $f \in \mathcal{P}\mathcal{L}$ , and for each  $s$ ,  $f(s, \cdot)$  is constant. Furthermore,  $f$  vanishes on the eight points  $y_v, z_v$ . As a consequence  $\Delta_{r+i}(f) = 0$  for  $1 \leq i \leq 4$ . For  $0 \leq i \leq r$ ,  $\Delta_i(f) = \Psi_i(u)$ . Since  $\Psi_i$  measures jumps in derivatives at points of  $P(\mathcal{N}_i)$ , we have  $\Psi_i(u) = 0$  for all  $i$  (except  $i=0$  and  $i=j$ ) in the range  $0 \leq i \leq r$ . Thus after computing we have

$$0 = \sum b_i \Delta_i(f) = b_0 \Psi_0(u) + b_j \Psi_j(u) = (a-c)^{-1} (b_j - b_0).$$

Case 2. We assume that  $a < c < b$ . Define

$$f(s, t) = u(s) = \begin{cases} 0, & s \leq a \\ (c-a)^{-1}(s-a), & a \leq s \leq c \\ -(b-c)^{-1}(s-b), & c \leq s \leq b \\ 0, & s \geq b. \end{cases}$$

Proceeding as before we arrive at

$$0 = b_0 \Psi_0(u) + b_j \Psi_j(u) = [(b - c)^{-1} + (c - a)^{-1}](b_0 - b_j).$$

Case 3.  $b < c$ . The calculations are like those in Case 1. ■

5.8. LEMMA. Let  $r + 1 = n$  and  $\sum_{i=0}^{r+4} b_i \Delta_i = 0$ . Then for  $i = 0, 1, \dots, r$ ,

$$b_i = \lambda_i b_{\varepsilon_4} + (1 - \lambda_i) b_{\varepsilon_2},$$

where  $\lambda_i = (\sigma_m - \sigma) / (\sigma_m - \sigma_1)$  and  $\sigma \in P(\mathcal{N}_i)$ .

*Proof.* If  $i = \varepsilon_4$  or  $i = \varepsilon_2$  the formula is trivial. We therefore assume  $i \neq \varepsilon_4$  and  $i \neq \varepsilon_2$ . Then  $\sigma \neq \sigma_1$  and  $\sigma \neq \sigma_m$ . Construct a function  $u \in \mathcal{PL}(\mathbb{R})$  as shown in Fig. 5.3. Let  $f(s, t) = u(s)$ . Then

$$0 = \sum_{j=0}^{r+4} b_j \Delta_j(f) = b_{\varepsilon_2} \frac{1}{\sigma_m - \sigma} + b_{\varepsilon_4} \frac{1}{\sigma - \sigma_1} - b_i \left( \frac{1}{\sigma_m - \sigma} + \frac{1}{\sigma - \sigma_1} \right).$$

Consequently

$$b_{\varepsilon_2}(\sigma - \sigma_1) + b_{\varepsilon_4}(\sigma_m - \sigma) - b_i(\sigma_m - \sigma_1) = 0$$

whence

$$b_i = \lambda_i b_{\varepsilon_4} + (1 - \lambda_i) b_{\varepsilon_2}. \quad \blacksquare$$

5.9. LEMMA. Let  $r + 1 = n$  and  $\sum_{i=0}^{r+4} b_i \Delta_i = 0$ . Then  $b_0 = b_1 = \dots = b_r$ .

*Proof.* From 5.7,  $b_i = \lambda_i b_{\varepsilon_4} + (1 - \lambda_i) b_{\varepsilon_2}$  where  $0 < \lambda_i < 1$ ,  $i = 0, \dots, r$ . Assume that  $b_{\varepsilon_4} \leq b_{\varepsilon_2}$ . From the equivalent lemma to 5.7 using the  $t$ -variable,

$$b_i = \mu_i b_{\varepsilon_1} + (1 - \mu_i) b_{\varepsilon_3}, \quad \text{where } 0 < \mu_i < 1, i = 0, 1, \dots, r.$$

From this we see that either  $\varepsilon_1 = \varepsilon_4$  and  $\varepsilon_2 = \varepsilon_3$  or  $\varepsilon_1 = \varepsilon_2$ ,  $\varepsilon_3 = \varepsilon_4$ .

Case (i).  $\varepsilon_1 = \varepsilon_4$  and  $\varepsilon_2 = \varepsilon_3$ . Then  $b_{r+1} + b_{r+2} = b_{r+3} + b_{r+4} = 0$  by 5.6, and so the previous lemma shows  $b_{\varepsilon_4} = b_{\varepsilon_2}$ . It then follows that  $b_{\varepsilon_4} = b_{\varepsilon_2} = b_i$ ,  $i = 0, \dots, r$ .

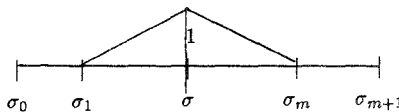


FIGURE 5.3



Case (ii).  $\varepsilon_1 = \varepsilon_2$  and  $\varepsilon_3 = \varepsilon_4$ . Then  $b_{r+2} + b_{r+3} = b_{r+1} + b_{r+4} = 0$  by 5.6, and so the previous lemma shows again that  $b_{\varepsilon_4} = b_{\varepsilon_2}$ . It then follows again that  $b_{\varepsilon_4} = b_{\varepsilon_2} = b_i$ ,  $i = 0, \dots, r$ . ■

5.10. THEOREM. *If  $\mathcal{N}$  contains no closed path, then the space  $\mathcal{RB}^\perp$  is spanned by the set  $\{\Delta_0, \dots, \Delta_{r+4}\}$ . The only dependence among these functionals, aside from a scalar multiple, is*

$$\sum_{i=0}^r \Delta_i + \lambda \sum_{i=1}^4 (-1)^{i+1} \Delta_{r+i} = 0.$$

*Proof.* If  $\sum_0^{r+4} b_i \Delta_i = 0$ , then by 5.7 and 5.9,  $b_0 = b_1 = \dots = b_r$ . By 5.5,  $b_{r+v} = (-1)^v \lambda b_0$  for  $v = 1, \dots, 4$ . This proves the second assertion of the theorem and that  $\{\Delta_0, \dots, \Delta_{r+4}\}$  spans a space of dimension  $r+4$ . This space is in  $\mathcal{RB}^\perp$  by 3.5 and 3.6. By 2.3, we have

$$n + r + 4 = \dim \mathcal{PL} = \dim \mathcal{RB} + \dim \mathcal{RB}^\perp = n + \dim \mathcal{RB}^\perp$$

and so  $\dim \mathcal{RB}^\perp = r + 4$ . ■

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