Contents

1	Der	ivatives for RHS of Weight System
	1.1	$\frac{\partial}{\partial x}$
	1.2	$\frac{\partial}{\partial y}$, and higher dimensions
		Laplacian ∇^2
		1.3.1 2D
		1.3.2 3D
		Projected Operators
	1.5	Problems with Paper

Chapter 1

Derivatives for RHS of Weight System

We start with:

$$\psi(x,y) = e^{-\epsilon^2(x^2 + y^2)} \cdot G_k(2\epsilon^2(xx_i + yy_i))$$
(1.1)

Fornberg et al. [?] provide that derivatives of the basis functions requires:

$$\frac{\partial^p G_k(z)}{\partial z^p} = G_{\max(0,k-p)}(z), \qquad p = 0, 1, 2 \cdots$$
 (1.2)

and derivatives of Equation 1.1 can be obtained via the product rule.

For RBF-GA, we have:

$$z = 2\epsilon^2(xx_i + yy_i) \tag{1.3}$$

1.1 $\frac{\partial}{\partial x}$

First, we test derivation for $\frac{\partial}{\partial x}$.

Apply the product rule:

$$\frac{\partial \psi(x,y)}{\partial x} = \frac{\partial}{\partial x} \left(e^{-\epsilon^2 (x^2 + y^2)} \cdot G_k(z) \right)
= \frac{\partial}{\partial x} \left(e^{-\epsilon^2 (x^2 + y^2)} \right) G_k(z) + e^{-\epsilon^2 (x^2 + y^2)} \frac{\partial G_k(z)}{\partial x}$$
(1.4)

Use the chain rule as necessary:

$$\frac{\partial}{\partial x} \left(e^{-\epsilon^2 (x^2 + y^2)} \right) = e^{-\epsilon^2 (x^2 + y^2)} \cdot \frac{\partial}{\partial x} \left(\epsilon^2 (x^2 + y^2) \right)$$

$$= e^{-\epsilon^2 (x^2 + y^2)} \left(2x\epsilon^2 \right) \tag{1.5}$$

$$\frac{\partial G_k(z)}{\partial x} = \frac{\partial z}{\partial x} \frac{\partial}{\partial z} G_k(z)$$

$$= \frac{\partial}{\partial x} \left(2\epsilon^2 (xx_i + yy_i) \right) G_{\max(0,k-p)}(z)$$

$$= 2\epsilon^2 x_i G_{\max(0,k-p)}(z) \tag{1.6}$$

Note that in the derivative of z we assume the center coordinates x_i and y_i are considered scalars.

Finally, we substitute Equation 1.5 and 1.6 into Equation 1.4 to get:

$$\frac{\partial \psi(x,y)}{\partial x} = \left(2x\epsilon^2 G_k(z) + \frac{\partial G_k(z)}{\partial x}\right) e^{-\epsilon^2(x^2+y^2)}$$

$$= \left(2x\epsilon^2 G_k(z) + 2\epsilon^2 x_i G_{\max(0,k-p)}(z)\right) e^{-\epsilon^2(x^2+y^2)}$$

$$= \left(xG_k(z) + x_i G_{\max(0,k-p)}(z)\right) 2\epsilon^2 e^{-\epsilon^2(x^2+y^2)}$$
(1.7)

1.2 $\frac{\partial}{\partial u}$, and higher dimensions

We repeat the process for $\frac{\partial}{\partial y}$, but due to similarity to $\frac{\partial}{\partial x}$, we can conclude:

$$\frac{\partial \psi(x,y)}{\partial y} = \left(y G_k(z) + y_i G_{\max(0,k-p)}(z) \right) 2\epsilon^2 e^{-\epsilon^2 (x^2 + y^2)} \tag{1.8}$$

As the problem dimension increases, each first order derivative results similar equations, but a substitution for $r(\mathbf{x})^2 = (x^2 + y^2 + \cdots)$ can be made:

$$\frac{\partial \psi(\mathbf{x})}{\partial x} = \left(x G_k(z) + x_i G_{\max(0,k-p)}(z) \right) 2\epsilon^2 e^{-\epsilon^2 r(\mathbf{x})^2}$$
(1.9)

1.3 Laplacian ∇^2

The Laplacian, $\nabla^2 = \sum_{d=1}^D \frac{\partial^2}{\partial x_d^2}$ (where D is the problem dimension) can be obtained following [?] to rewrite Equation 1.1 as the product of two functions in one variable.

- 1.3.1 2D
- 1.3.2 3D

1.4 Projected Operators

For the Stokes problem we need projected operators on the sphere. Derivation of these operators can be done via the chain rule, but I suspect that we may be able to compose the projected operators using only the basic operators for $\frac{\partial}{\partial x}$, $\frac{\partial}{\partial y}$, and $\frac{\partial}{\partial z}$ as demonstrated in one of my dissertation Appendices—I was able to compose Differentiation Matrices for projected operators based on DMs for the unprojected operators with minimal loss of accuracy.

In theory the composition/indirect approach to obtaining the projected operators will suffice. If not, we shall return to derive the RHS forms.

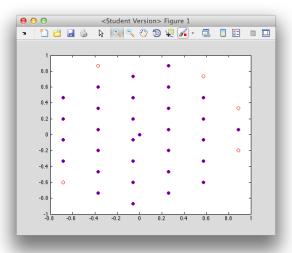


Figure 1.1: Red: A hexagonal grid within the unit disk $(r \le 1)$. Blue: a stencil of size n = 31 nodes selected from the grid.

1.5 Problems with Paper

- In Equations 12-14 in Bengt's paper, the growth rate for the power on the $\frac{1}{\epsilon^*}$ is not obvious. Through trial and error I was able to deduce that the power should scale as: $\frac{1}{\epsilon^{2k}}$ where k is the same k on G_k .
- When $z = 2\epsilon^2 x \cdot x_i$ is negative (i.e., the cases where stencil nodes are outside of first quadrant), the gammainc function in matlab returns NaN values. How do we handle the situation?
 - Actually, the problem is not when they are outside the first quadrant. Instead, it appears when the nodes span multiple quadrants (e.g., when in $[-1,1]^d$ instead of $[0,1]^d$).
 - Answer: The imag(gammainc(z,k)) < 1e-15 for all cases tested. . Since it is so near to 0, we simply exclude it by requesting the real(gammainc(z,k)) when producing the bases.
- I am attempting to generate figures similar to Figure 3. I am able to reproduce the basis functions for $\Psi_1(x), \Psi_2(x), \Psi_3(x)$, but the remainder do not reproduce the figures in the paper.
 - Note: I was playing with different node distributions (regular, hexagonal, halton), and I realized that the visualized bases depend on the stencil and node locations.
 Bengt's paper does not explicitly state what node distribution was used to produce that figure. But at least I am confident that the results in Figure 1.2 (and the X-Y axes view in Figure 1.3) are correct for my node distributions.

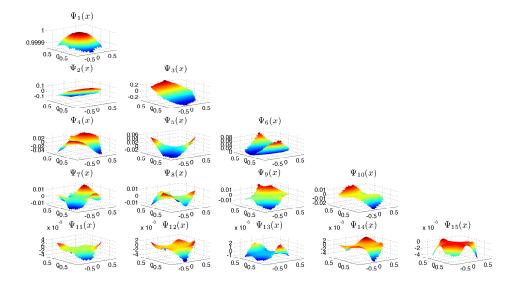


Figure 1.2: 3D view of the new basis functions from RBF-GA for the hexagonal grid and stencil shown in Figure 1.1 and stencil size n = 31.

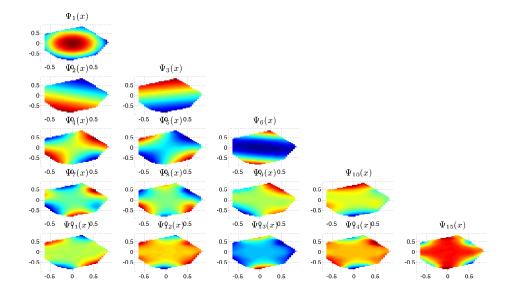


Figure 1.3: 2D view of the new basis functions from RBF-GA for the hexagonal grid and stencil shown in Figure 1.1 and stencil size n = 31.