# Proof of the Second Partials Test Edward Burkard

### 1. PROOF OF THE SECOND DERIVATIVE TEST FROM CALC I (USING CALC II)

Recall from Calc II that the Taylor polynomial of a function f at a point a is given by

$$f(x) = f(a) + f'(a)(x - a) + \frac{1}{2}f''(a)(x - a)^{2} + \cdots$$

If a is a critical point of f, that is f'(a) = 0, then the Taylor polynomial reads

$$f(x) = f(a) + \frac{1}{2}f''(a)(x-a)^2 + \cdots$$

This means that when x is very close to a (so that the quantity (x - a) is small), the behavior of f is near completely determined by the second order term:  $\frac{1}{2}f''(a)(x-a)^2$ . This is because  $(x-a)^2 >> (x-a)^3$ when (x - a) is very small. So, we see that if f''(a) > 0, then near a, f looks like a parabola opening upward, hence a is a local minimum; and if f''(a) < 0, then f looks like a parabola opening downward, hence a is a local maximum. Here is a graph of the function f(x) = 8x(x-1)(x+1):



Notice how, inside the blue box (which contains the local maximum of f(x)), the function looks almost like (but not exactly like) a parabola that opens downward; and inside the purple box (which contains the local minimum of f(x)), the function looks almost like a parabola that opens upward.

Now, if the second derivative is also zero at a, f''(a) = 0, but the third derivative is not zero,  $f'''(a) \neq 0$ , then the Taylor series is dominated by the third order term:  $\frac{1}{6}f'''(a)(x-a)^3$ . This explains why the second derivative being zero gives a point of inflection. We can continue onward like this for as long as necessary (e.g., if f'''(a) = 0, then move on to  $f^{(4)}(a)$  so that the behavior of f near a is near completely determined by the fourth order term  $\frac{1}{24}f^{(4)}(a)(x-a)^4$ , etc...). This means that if f''(a) = 0, we don't have enough information to determine what type of critical point a is without taking more derivatives (hence the second derivative test fails).

# 2. PROOF OF THE SECOND PARTIALS TEST

To prove the second partials test, we are going to try to mimic the above proof in the one variable case.

2.1. Multiplying a vector by a matrix. To make our lives easier, we should think of vectors as columns, e.g., instead of writing the vector  $\vec{v} = \langle h, k \rangle$ , we write  $\vec{v} = \begin{pmatrix} h \\ k \end{pmatrix}$ . This will make our dealing with the following proof a bit easier. Suppose we have an  $2 \times 2$  matrix

$$A = \left(\begin{array}{cc} a_{11} & a_{12} \\ a_{21} & a_{22} \end{array}\right),$$

Then the product  $A\vec{v}$  is the vector given by

$$A\vec{v} = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} \begin{pmatrix} h \\ k \end{pmatrix} = \begin{pmatrix} a_{11}h + a_{12}k \\ a_{12}h + a_{22}k \end{pmatrix}$$

2.2. Second order Taylor polynomial of f(x, y). We will gloss over some technicalities here, but they can be found in section 4.1 of Susan Colley's book "Vector Calculus" [1]. The second order Taylor polynomial of a  $C^2$  (continuous second partials) f(x, y) about a point A = (p, q) is given by

$$f(x,y) = f(p,q) + \nabla f(p,q) \cdot (\vec{x} - \vec{a}) + \frac{1}{2}(\vec{x} - \vec{a}) \cdot (Hf(p,q)(\vec{x} - \vec{a})) + R_2(\vec{x},\vec{a})$$

where  $\vec{x} = \begin{pmatrix} x \\ y \end{pmatrix}$  is the position vector of an arbitrary point (x, y),  $\vec{a} = \begin{pmatrix} p \\ q \end{pmatrix}$  is the position vector of A, Hf(p,q) is the Hessian of f at A, and  $R_2(\vec{x}, \vec{a})$  is the remainder term which satisfies

$$\frac{|R_2(\vec{x}, \vec{a})|}{|\vec{x} - \vec{a}|^2} \to 0 \qquad \text{as} \qquad \vec{x} \to \vec{a}.$$

The dot,  $\cdot$ , is the usual dot product of vectors. If we write this out, we have

$$\begin{aligned} f(x,y) &= f(p,q) + \begin{pmatrix} f_x(p,q) \\ f_y(p,q) \end{pmatrix} \cdot \begin{pmatrix} x-p \\ y-q \end{pmatrix} + \frac{1}{2} \begin{pmatrix} x-p \\ y-q \end{pmatrix} \cdot \left[ \begin{pmatrix} f_{xx}(p,q) & f_{xy}(p,q) \\ f_{yx}(p,q) & f_{yy}(p,q) \end{pmatrix} \begin{pmatrix} x-p \\ y-q \end{pmatrix} \right] \\ &= f(p,q) + \begin{pmatrix} f_x(p,q) \\ f_y(p,q) \end{pmatrix} \cdot \begin{pmatrix} x-p \\ y-q \end{pmatrix} + \frac{1}{2} \begin{pmatrix} x-p \\ y-q \end{pmatrix} \cdot \begin{pmatrix} f_{xx}(p,q)(x-p) + f_{xy}(p,q)(y-q) \\ f_{xy}(p,q)(x-p) + f_{yy}(p,q)(y-q) \end{pmatrix} \\ &= f(p,q) + f_x(p,q)(x-p) + f_y(p,q)(y-q) \\ &+ \frac{1}{2} \left[ f_{xx}(p,q)(x-p)^2 + 2f_{xy}(p,q)(x-p)(y-q) + f_{yy}(p,q)(y-q)^2 \right] \end{aligned}$$

Just as in Calc II, this series has a radius of convergence R which it is valid in (it gives a disk of points about the point A of radius R which the series is valid in).

With this, we can quantify the change in f between the point (p,q) and some point (x,y), which is given by

$$\Delta f = f(x, y) - f(p, q)$$

as

$$\Delta f = f_x(p,q)(x-p) + f_y(p,q)(y-q) + \frac{1}{2} \left[ f_{xx}(p,q)(x-p)^2 + 2f_{xy}(p,q)(x-p)(y-q) + f_{yy}(p,q)(y-q)^2 \right]$$

2.3. A brief lemma. To prove the second derivative test, we use the following lemma:

**Lemma.** Consider the quadratic  $(A \neq 0)$  function  $g(x) = Ax^2 + 2Bx + C$ .

- (1) If  $AC B^2 > 0$ , and A > 0 or C > 0, then g(x) > 0 for all x.
- (2) If  $AC B^2 > 0$ , and A < 0 or C < 0, then g(x) < 0 for all x.
- (3) If  $AC B^2 < 0$ , then there are x values such that g(x) > 0 and some x values with g(x) < 0.

*Proof.* We prove these on a case by case basis

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(1) Suppose we have that  $AC - B^2 > 0$ . If A > 0, then  $\lim_{x \to \infty} g(x) = \infty$ , meaning that g(x) > 0 for some x. On the other hand, if C > 0 then g(0) > 0, so again, we know there are x where g(x) > 0. If g ever becomes negative, then by the intermediate value theorem, we know that g has zeros. We can use the quadratic formula to search for the x values for which g(x) = 0:

$$x = \frac{-2B \pm \sqrt{(2B)^2 - 4AC}}{2A} = \frac{-B \pm \sqrt{B^2 - AC}}{A}$$

Since  $AC - B^2 > 0$ , this means that  $B^2 - AC < 0$ , and so the x values from the quadratic formula above are not real (they have a nonzero imaginary part). This means that g(x) is never zero for any x, and so g never crosses below the x-axis, hence g(x) > 0 for all z.

(2) Almost identically to the previous part, suppose we have that  $AC - B^2 > 0$ . If A < 0, then  $\lim_{x \to \infty} g(x) = -\infty$ , meaning that g(x) < 0 for some x. On the other hand, if C < 0 then g(0) < 0, so again, we know there are x where g(x) < 0. If g ever becomes positive, then by the intermediate value theorem, we know that g has zeros. We can use the quadratic formula to search for the x values for which g(x) = 0:

$$z = \frac{-B \pm \sqrt{B^2 - AC}}{A}.$$

Since  $AC - B^2 > 0$ , this means that  $B^2 - AC < 0$ , and so the x values from the quadratic formula above are not real (they have a nonzero imaginary part). This means that g(x) is never zero for any x, and so g never crosses above the x-axis, hence g(x) < 0 for all z.

(3) Now, the fun part! Assume that  $AC - B^2 < 0$ . This means that  $B^2 - AC > 0$ . Let's search for when g(x) = 0. It is zero for the following z values:

$$x = \frac{-B \pm \sqrt{B^2 - AC}}{A}$$

Since  $B^2 - AC > 0$ , this means there are two real values of x for which g(x) is zero. Since g(x) has exactly two zeroes, it crosses the x-axis exactly twice. This must mean that g(x) takes on both negative and positive values. (If you're having trouble with this, just draw a few pictures of parabolas which intersect the x-axis twice to see it.)

#### 2.4. Proof of the test. Let us recall the theorem that we want to prove

**Theorem** (Second Partials Test). Suppose the second partial derivatives of f are continuous on a disk with center (a, b), and suppose that (a, b) is a critical point of f, i.e.,  $\nabla f(a, b) = \vec{0}$ . Let

$$D(a,b) = \det(Hf(a,b)),$$

then

- (1) If D(a,b) > 0 and  $f_{xx}(a,b) > 0$ , then f(a,b) is a local minimum.
- (2) If D(a,b) > 0 and  $f_{xx}(a,b) < 0$ , then f(a,b) is a local maximum.
- (3) If D(a,b) < 0, then (a,b) is a saddle point.

Recall that a point (a, b) is a

- (1) local minimum if  $\Delta f \ge 0$  for all (x, y) near (a, b),
- (2) local maximum if  $\Delta f \leq 0$  for all (x, y) near (a, b),
- (3) saddle point if  $\Delta f > 0$  for some (x, y) near (a, b) and  $\Delta f < 0$  for other (x, y) near (a, b).

Alright! Here we go!

*Proof.* Since (a, b) is a critical point, we know that  $f_x(a, b) = f_y(a, b) = 0$ , and so

$$\Delta f = \frac{1}{2} \left[ f_{xx}(a,b)(x-a)^2 + 2f_{xy}(a,b)(x-a)(y-b) + f_{yy}(a,b)(y-b)^2 \right]$$
$$= \frac{(y-b)^2}{2} \left[ f_{xx}(a,b) \left(\frac{x-a}{y-b}\right)^2 + 2f_{xy}(a,b) \left(\frac{x-a}{y-b}\right) + f_{yy}(a,b) \right]$$

Obviously, we avoid picking points where y = b for this proof, otherwise this will not work (we can pick points with y-coordinate as close to b as we want though). If we let  $z = \frac{x-z}{y-b}$ , then we have

$$\Delta f = \frac{(y-b)^2}{2}g(z)$$

where  $g(z) = f_{xx}(a,b)z^2 + 2f_{xy}(a,b)z + f_{yy}(a,b)$ . Since  $\frac{(y-b)^2}{2} \ge 0$ , the sign of  $\Delta f$  is completely determined by g(z). But g(z) is exactly of the form in the lemma above, where

$$\begin{array}{rcl}
A &=& f_{xx}(a,b) \\
B &=& f_{xy}(a,b) \\
C &=& f_{yy}(a,b) \\
AC - B^2 &=& \det(Hf(a,b)) = D(a,b)
\end{array}$$

- (1) Suppose that D(a,b) > 0 and  $f_{xx}(a,b) > 0$ . This means that  $AC B^2 > 0$  and A > 0 in terms of the lemma. In this case it means that g(z) > 0 for all z. Thus  $\Delta f$  is always positive, meaning that (a,b) is a local minimum.
- (2) Suppose that D(a,b) > 0 and  $f_{xx}(a,b) < 0$ . This means that  $AC B^2 > 0$  and A < 0 in terms of the lemma. In this case it means that g(z) < 0 for all z. Thus  $\Delta f$  is always negative, meaning that (a,b) is a local maximum.
- (3) Suppose that D(a,b) < 0. Then  $AC B^2 < 0$  and so g(z) is positive for some z, and negative for others. This means that  $\Delta f$  is positive for some points (x, y) and negative for others. Thus (a, b) is a saddle point.

A brief caveat for (1) and (2): Technically we have not shown that  $\Delta f \ge 0$  (resp.  $\Delta f \le 0$ ) for points (x, y) when y = b. To do this, in the equation for  $\Delta f$ , instead of factoring out  $(y - b)^2$ , we instead factor out  $(x - a)^2$ . In this case  $g(w) = f_{yy}(a, b)w^2 + 2f_{xy}(a, b)w + f_{xx}(a, b)$ , where  $w = \frac{y - b}{x - a}$ . This is why the conditions on C are in the lemma! This allows us to use the points when y = b (and disallows when x = a, but this was already taken care of in the previous case).

There is also the part that if D(a, b) = 0, then the test fails. This has to do with the nature of matrices. If the determinant of a matrix is zero, then that matrix is called *degenerate*. A degenerate matrix "maps one or more directions to zero". This usually corresponds to something where you have a whole line of critical points, e.g., the function  $f(x, y) = -(x - y)^2$ , or something more subtle, e.g.,  $f(x, y) = x^3 + y^3$ .

#### References

[1] Susan J. Colley, Vector Calculus, 4e. Pearson Education, Inc. Boston. 2012.