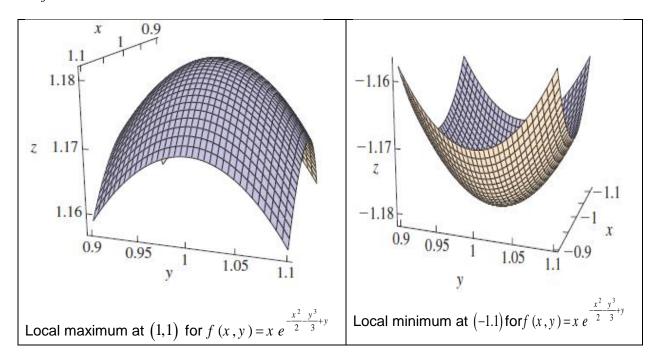
12.10 Extrema of functions of several variables

Definition1

We call f(a,b) a **local maximum** of f if there is an open disk R centered at (a,b), for which $f(a,b) \ge f(x,y)$ for all $(x,y) \in R$. Similarly, f(a,b) is called a **local minimum** of f if there is an open disk R centered at (a,b), for which $f(a,b) \le f(x,y)$ for all $(x,y) \in R$. In either case, f(a,b) is called a **local extremum** of f.



Definition2

The point (a,b) is a **critical point** of the function f(x,y) if (a,b) is in the domain of f and either $\frac{\partial f}{\partial x}(a,b) = \frac{\partial f}{\partial y}(a,b) = 0$ or one or both of $\frac{\partial f}{\partial x}$ and $\frac{\partial f}{\partial y}$ do not exist at (a,b).

Theorem1

 $\overline{\text{If } f(x,y)}$ has a local extremum at (a,b), then (a,b) must be a critical point of f.

Example1

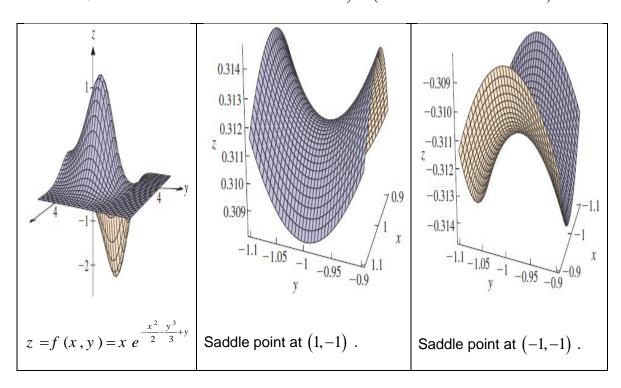
Find all critical points of $f(x,y) = x e^{-\frac{x^2}{2} - \frac{y^3}{3} + y}$.

Solution

First, we compute the first partial derivatives:

$$\frac{\partial f}{\partial x}(x,y) = (1-x^2)e^{-\frac{x^2}{2}-\frac{y^3}{3}+y} \text{ and } \frac{\partial f}{\partial y}(x,y) = x(1-y^2)e^{-\frac{x^2}{2}-\frac{y^3}{3}+y}.$$

Since exponentials are always positive, we have $\frac{\partial f}{\partial x}(x,y)=0$ if and only if $1-x^2=0$, that is, when $x=\pm 1$. We have $\frac{\partial f}{\partial y}(x,y)=0$ if and only if $x\left(1-y^2\right)=0$, that is, when x=0 or $y=\pm 1$. So the set of critical points is $C_f=\left\{(-1,-1),(-1,1),(1,-1),(1,1)\right\}$.



Definition3

The point P(a,b,f(a,b)) is a **saddle point** of z = f(x,y) if (a,b) is a critical point of f and if every open disk centered at (a,b) contains points (x,y) in the domain of f for which f(x,y) < f(a,b) and points (x,y) in the domain of f for which f(x,y) > f(a,b).

Theorem2 (Second Derivatives Test)

Suppose that $f\left(x\,,y\,\right)$ has continuous second-order partial derivatives in some open disk containing the point (a,b) and that $f_{x}\left(a,b\right)=f_{y}\left(a,b\right)=0$. Define the discriminant D for the point (a,b) by $D(a,b)=f_{xx}\left(a,b\right)f_{yy}\left(a,b\right)-\left[f_{xy}\left(a,b\right)\right]^{2}$.

- If D(a,b) > 0 and $f_{xx}(a,b) > 0$, then f has a local minimum at (a,b).
- If D(a,b) > 0 and $f_{xx}(a,b) < 0$, then f has a local maximum at (a,b).
- If D(a,b) < 0, then f has a saddle point at (a,b).
- If D(a,b) = 0, then no conclusion can be drawn.

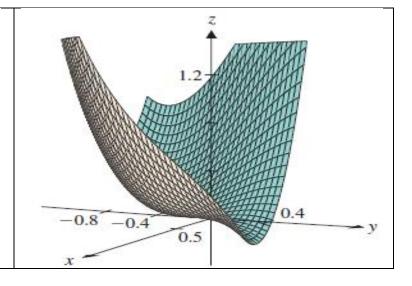
Example2 (Using the Discriminant to Find Local Extrema)

Locate and classify all critical points for $f(x,y) = 2x^2 - y^3 - 2xy$.

Solution

We first compute the first partial derivatives: $f_x=4x-2y$ and $f_y=-3y^2-2x$. Since both f_x and f_y are defined for all $\left(x,y\right)$, the critical points are solutions of the two equations: $f_x=4x-2y=0$ and $f_y=-3y^2-2x=0$. Solving the first equation for y, we get y=2x. Substituting this into the second equation, we have $0=-3(4x^2)-2x=-12x^2-2x=-2x(6x+1)$, so that x=0 or $x=\frac{-1}{6}$. The corresponding y-values are y=0 and $y=\frac{-1}{3}$. The only two critical points are then (0,0) and $\left(\frac{-1}{6},\frac{-1}{3}\right)$. To classify these points, we first compute the second partial derivatives: $f_{xx}=4$, $f_{yy}=-6y$ and $f_{xy}=-2$, and then test the discriminant. We have $D(0,0)=4\times0-(-2)^2=-4<0$ and $D\left(\frac{-1}{6},\frac{-1}{3}\right)=4\times(-6)\times\left(\frac{-1}{3}\right)-(-2)^2=4>0$. From Theorem 2, we conclude that there is a <u>saddle point</u> of f at (0,0), since D(0,0)<0. Further, there is a <u>local minimum at $\left(\frac{-1}{6},\frac{-1}{3}\right)$ since $D\left(\frac{-1}{6},\frac{-1}{3}\right)>0$ and $f_{xx}\left(\frac{-1}{6},\frac{-1}{3}\right)=4>0$.</u>

Point	(0, 0)	$\left(-\frac{1}{6}, -\frac{1}{3}\right)$
$f_{xx}=4$	4	4
$f_{yy} = -6y$	0	2
$f_{xy} = -2$	-2	-2
D(a, b)	-4	4



Example3 (Classifying Critical Points)

Locate and classify all critical points for $f(x,y) = x^3 - 2y^2 - 2y^4 + 3x^2y$.

Solution

Here, we have $f_x=3x^2+6xy$ and $f_y=-4y-8y^3+3x^2$. Since both f_x and f_y exist for all (x,y), the critical points are solutions of the two equations: $f_x=3x^2+6xy=0$ and $f_y=-4y-8y^3+3x^2=0$. From the first equation, we have $0=3x^2+6xy=3x(x+2y)$, so that at a critical point, x=0 or x=-2y.

Substituting x=0 into the second equation, we have $0=-4y-8y^3=-4y(1+2y^2)$. The only (real) solution of this equation is y=0. This says that for x=0, we have only one critical point: (0,0).

Substituting x = -2y into the second equation, we have

 $0 = -4y - 8y^3 + 3(-2y)^2 = -4y(1 + 2y^2 - 3y) = -4y(2y - 1)(y - 1)$. The solutions of this equation are y = 0, $y = \frac{1}{2}$ and y = 1, with corresponding critical points (0,0), $(-1,\frac{1}{2})$ and (-2,1).

To classify the critical points, we compute the second partial derivatives,

$$f_{xx} = \frac{\partial}{\partial x} \left(3x^2 + 6xy \right) = 6x + 6y \quad f_{yy} = \frac{\partial}{\partial y} \left(-4y - 8y^3 + 3x^2 \right) = -4 - 24y^2 \text{ , and }$$

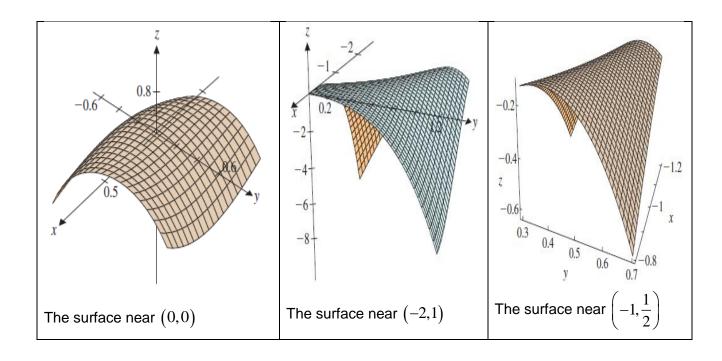
 $f_{xy} = \frac{\partial}{\partial y} (3x^2 + 6xy) = 6x$, and evaluate the discriminant at each critical point. We

have D(0,0)=0, $D\left(-1,\frac{1}{2}\right)=-6<0$ and D(-2,1)=24>0 . From Theorem 2, we

conclude that \underline{f} has a saddle point at $\left(-1, \frac{1}{2}\right)$, since $D\left(-1, \frac{1}{2}\right) = -6 < 0$. Further, \underline{f} has a

<u>local maximum at</u> (-2,1) since D(-2,1)=24>0 and $f_{xx}(-2,1)=-3<0$. Unfortunately, Theorem 2 gives us no information about the critical point (0,0), since D(0,0)=0.

However, notice that in the plane y=0 we have $f(x,y)=x^3$. In two dimensions, the curve $z=x^3$ has an inflection point at x=0. This shows that there is <u>no local extremum at (0,0)</u>.



Definition4

We call f(a,b) the **absolute maximum** of f on the region R if $f(a,b) \ge f(x,y)$ for all $(x,y) \in R$. Similarly, f(a,b) is called the **absolute minimum** of f on R if $f(a,b) \le f(x,y)$ for all $(x,y) \in R$. In either case, f(a,b) is called an **absolute** extremum of f.

Theorem 3 (Extreme Value Theorem)

Suppose that f(x,y) is continuous on the closed and bounded region $R \subset \mathbb{R}^2$. Then f has both an absolute maximum and an absolute minimum on R. Further, an absolute extremum may only occur at a critical point in R or at a point on the boundary of R.

12.11 Constrained Optimization and Lagrange Multipliers

In this section, we develop a technique for finding the maximum or minimum of a function, given one or more constraints on the function's domain.

Theorem1

Suppose that f(x,y,z) and g(x,y,z) are functions with continuous first partial derivatives and $\nabla g(x,y,z) \neq 0$ on the surface g(x,y,z) = 0. Suppose that either the minimum (or the maximum) value of f(x,y,z) subject to the constraint g(x,y,z) = 0 occurs at $\left(x_0,y_0,z_0\right)$. Then $\nabla f\left(x_0,y_0,z_0\right) = \lambda \nabla g\left(x_0,y_0,z_0\right)$, for some constant λ (called a Lagrange multiplier).

Remark1

• Note that Theorem 1 says that if f(x,y,z) has an extremum at a point (x_0,y_0,z_0) on the surface g(x,y,z)=0, we will have for $(x,y,z)=(x_0,y_0,z_0)$,

$$\begin{cases} f_x(x, y, z) = \lambda g_x(x, y, z) \\ f_y(x, y, z) = \lambda g_y(x, y, z) \\ f_z(x, y, z) = \lambda g_z(x, y, z) \\ g(x, y, z) = 0 \end{cases}$$

Finding such extrema then boils down to solving these four equations for the four unknowns x, y, z and λ .

• Notice that the Lagrange multiplier method we have just developed can also be applied to functions of two variables, by ignoring the third variable in Theorem1. That is, if f(x,y) and g(x,y) have continuous first partial derivatives and $f\left(x_{0},y_{0}\right)$ is an extremum of f, subject to the constraint g(x,y)=0, then we must have $\nabla f\left(x_{0},y_{0}\right)=\lambda\nabla g\left(x_{0},y_{0}\right)$, for some constant λ . In this case, we end up with the three equations $f_{x}(x,y)=\lambda g_{x}(x,y)$, $f_{y}(x,y)=\lambda g_{y}(x,y)$ and g(x,y)=0, for the three unknowns x,y and λ .

Example 1 (Finding a Minimum Distance)

Use Lagrange multipliers to find the point on the line y = 3 - 2x that is closest to the origin.

Solution

For $f(x,y) = x^2 + y^2$, we have $\nabla f(x,y) = \langle 2x, 2y \rangle$ and for g(x,y) = 2x + y - 3, we have $\nabla g(x,y) = \langle 2,1 \rangle$. The vector equation $\nabla f(x,y) = \lambda \nabla g(x,y)$ becomes $\langle 2x, 2y \rangle = \lambda \langle 2, 1 \rangle$ from which it follows that $2x = 2\lambda$ and $2y = \lambda$.

The second equation gives us $\lambda = 2y$. The first equation then gives us $x = \lambda = 2y$. Substituting x = 2y into the constraint equation y = 3 - 2x, we have 5y = 3.

The solution is
$$y = \frac{3}{5}$$
, giving us $x = 2y = \frac{6}{5}$. The closest point is then $\left(\frac{6}{5}, \frac{3}{5}\right)$.

Example 2 (Optimization with an Inequality Constraint)

Suppose that the temperature of a metal plate is given by $T(x,y) = x^2 + 2x + y^2$, for points (x,y) on the elliptical plate defined by $x^2 + 4y^2 \le 24$. Find the maximum and minimum temperatures on the plate.

Solution

The plate corresponds to the shaded region R shown in Figure 1.

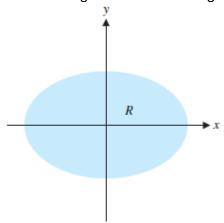


Figure 1: A metal plate.

We first look for critical points of T(x,y) inside the region R. We have $\nabla T(x,y) = <2x+2$, 2y>=<0, 0> if (x,y)=(-1,0), which is in R. At this point, T(-1,0)=-1. We next look for the extrema of T(x,y) on the ellipse $x^2+4y^2=24$. We first rewrite the constraint equation as $g(x,y)=x^2+4y^2-24=0$. From Theorem 1, any extrema on the ellipse will satisfy the Lagrange multiplier equation: $\nabla T(x,y)=\lambda \nabla g(x,y)$ or <2x+2, $2y>=\lambda <2x$, $8y>=<2\lambda x$, $8\lambda y>$. This occurs when $2x+2=2\lambda x$ and $2y=8\lambda y$.

Notice that the second equation holds when y = 0 or $\lambda = \frac{1}{4}$.

If y = 0, the constraint $x^2 + 4y^2 = 24$ gives $x = \pm \sqrt{24}$.

If $\lambda = \frac{1}{4}$, the first equation becomes $2x + 2 = \frac{1}{2}x$ so that $x = -\frac{4}{3}$. The constraint

$$x^2 + 4y^2 = 24$$
 now gives $y = \pm \frac{\sqrt{50}}{3}$.

Finally, we compare the function values at all of these points (the one interior critical point and the candidates for boundary extrema):

and
$$T(-1,0) = -1$$
, $T(\sqrt{24},0) = 24 + \sqrt{24} \approx 33.8$, $T(-\sqrt{24},0) = 24 - 2\sqrt{24} \approx 14.2$
 $T\left(-\frac{4}{3}, \frac{\sqrt{50}}{3}\right) = \frac{14}{3} \approx 4.7$, $T\left(-\frac{4}{3}, -\frac{\sqrt{50}}{3}\right) = \frac{14}{3} \approx 4.7$.

From this list, it's easy to identify the minimum value of -1 at the point $\left(-1,0\right)$ and the maximum value of $24+2\sqrt{24}$ at the point $\left(\sqrt{24},0\right)$.

We close this section by considering the case of finding the minimum or maximum value of a differentiable function f(x, y, z) subject to two constraints g(x, y, z) = 0 and h(x, y, z) = 0, where g and h are also differentiable (see Figure 2 below).

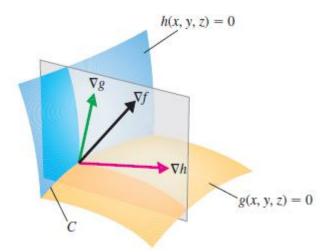


Figure 2: Constraint surfaces and the plane determined by the normal vectors ∇g and ∇h .

The method of Lagrange multipliers for the case of two constraints then consists of finding the point (x, y, z) and the Lagrange multipliers λ and μ (for a total of five unknowns) satisfying the five equations defined by:

$$\begin{cases} f_x(x, y, z) = \lambda g_x(x, y, z) + \mu h_x(x, y, z) \\ f_y(x, y, z) = \lambda g_y(x, y, z) + \mu h_y(x, y, z) \\ f_z(x, y, z) = \lambda g_z(x, y, z) + \mu h_z(x, y, z) \\ g(x, y, z) = 0 & & h(x, y, z) = 0 \end{cases}$$

Example 3 (Optimization with Two Constraints)

The plane x + y + z = 12 intersects the paraboloid $z = x^2 + y^2$ in an ellipse. Find the point on the ellipse that is closest to the origin.

Solution

We illustrate the intersection of the plane with the paraboloid in Figure 3.

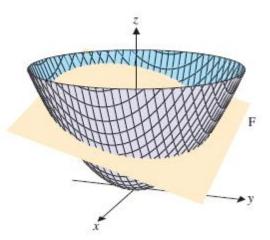


Figure 3: Intersection of a paraboloid and a plane.

Observe that minimizing the distance to the origin is equivalent to minimizing $f(x,y,z)=x^2+y^2+z^2$ [the *square* of the distance from the point (x,y,z) to the origin]. Further, the constraints may be written as g(x,y,z)=x+y+z-12=0 and $h(x,y,z)=x^2+y^2-z=0$. At any extremum, we must have that

$$\nabla f(x, y, z) = \lambda \nabla g(x, y, z) + \mu \nabla h(x, y, z)$$
 or $< 2x, 2y, 2z > = \lambda < 1, 1, 1 > + \mu < 2x, 2y, -1 > .$

Together with the constraint equations, we now have the system of equations:

$$\begin{cases} 2x = \lambda + 2\mu x & (1) \\ 2y = \lambda + 2\mu y & (2) \\ 2z = \lambda - \mu & (3) \\ x + y + z - 12 = 0 & (4) & & x^2 + y^2 - z = 0 & (5) \end{cases}$$

From (1), we have $\lambda = 2x(1-\mu)$, while from (2), we have $\lambda = 2y(1-\mu)$.

Setting these two expressions for λ equal gives us $2x(1-\mu) = 2y(1-\mu)$,

from which it follows that either $\mu = 1$ (in which case $\lambda = 0$) or x = y. However, if $\mu = 1$ and $\lambda = 0$, we have from (3) that z = -12, which contradicts (5).

Consequently, the only possibility is to have x = y, from which it follows from (5) that $z = 2x^2$. Substituting this into (4) gives us:

 $0 = x + y + z - 12 = x + x + 2x^2 - 12 = 2x^2 + 2x - 12 = 2(x + 3)(x - 2)$, so that x = -3 or x = 2. Since y = x and $z = 2x^2$, we have that (2, 2, 8) and (-3, -3, 18) are the

only candidates for extrema. Finally, since f(2,2,8) = 72 and f(-3,-3,18) = 342,

the closest point on the intersection of the two surfaces to the origin is (2,2,8). By the same reasoning, observe that the farthest point on the intersection of the two surfaces from the origin is (-3,-3,18).