

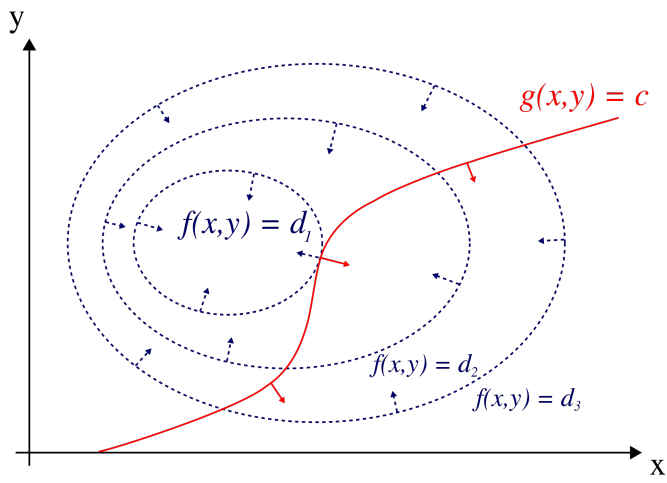
12.9 Lagrange Multiplier Thm to solve constrained optimization problem:

If a maximum or minimum of

$$z = f(x, y) \text{ subject to the constraint } g(x, y) = 0$$

occurs at point \mathbf{p} , then either

- 1.) $\nabla g(\mathbf{p}) = 0$
- 2.) There exists a constant λ such that $\nabla f(\mathbf{p}) = \lambda \nabla g(\mathbf{p})$



Thus to find extrema for constrained optimization problem,

- 1.) Find all \mathbf{p} that satisfy $\nabla g(\mathbf{p}) = 0$.
- 2.) Find all \mathbf{p} that satisfy $\nabla f(\mathbf{p}) = \lambda \nabla g(\mathbf{p})$ for some constant λ .
- 3.) Check if the points found in steps 1 and 2 satisfy the constraint $g(x, y) = 0$

Example: Find maximum $z = -x^2 - y^2 + 25$ subject to the constraint $xy - 1 = 0$.

Note $f(x, y) = -x^2 - y^2 + 25$ and $g(x, y) = xy - 1 = 0$

Easier method 1: Since one can solve $g(x, y) = 0$ for y (or x), one can use constraint to solve for y (or x) and plug into f , and then use (1) algebra or (2) calc 1 or (3) section 12.5

Method 2: Lagrange multiplier

Note: normally only use Lagrange multiplier when can't (or don't want to) solve constraint for y (or x).

$$\nabla g = \langle y, x \rangle$$

If $\nabla g = \langle y, x \rangle = \langle 0, 0 \rangle$, then $(x, y) = (0, 0)$, but $g(0, 0) = 0 - 1 \neq 0$. Thus $(0, 0)$ does not satisfy the constraint.

$$\nabla f = \langle -2x, -2y \rangle$$

$$\text{Solve } \langle -2x, -2y \rangle = \lambda \langle y, x \rangle$$

$$-2x = \lambda y \quad \text{and} \quad -2y = \lambda x$$

$$\text{Thus } -2x^2 = \lambda xy = -2y^2$$

$$\text{Thus } x^2 = y^2$$

$$\text{Thus } (x, y) = (1, 1), (1, -1), (-1, 1), (-1, -1)$$

BUT only 2 of these points satisfy the constraint $g(x, y) = xy - 1 = 0$

$$g(1, 1) = 1 - 1 = 0$$

$$g(1, -1) = -1 - 1 \neq 0$$

$$g(-1, -1) = 1 - 1 = 0$$

$$g(-1, 1) = -1 - 1 \neq 0$$

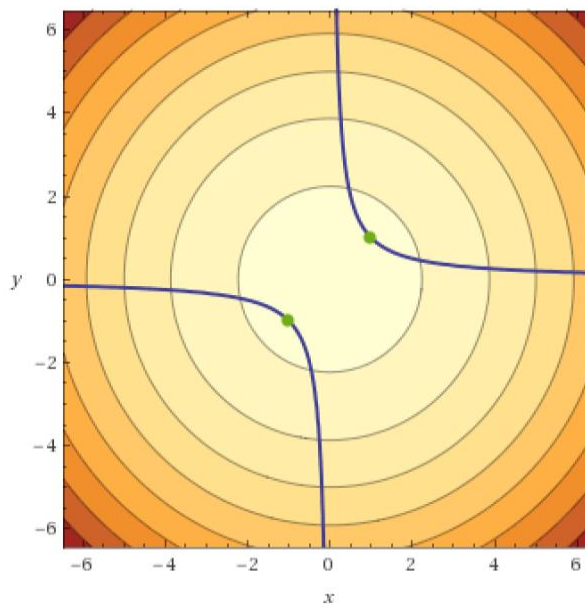
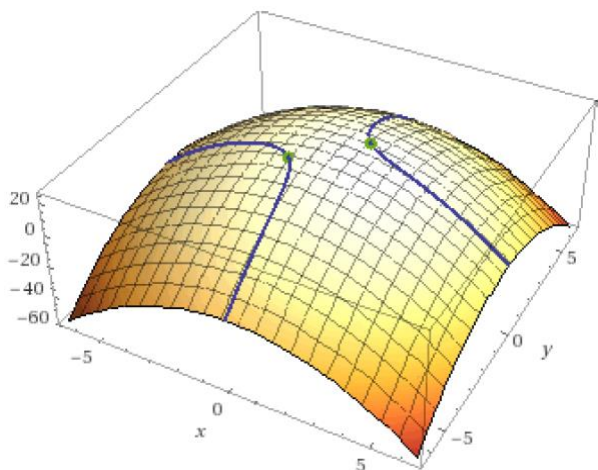
Thus only $(1, 1)$ and $(-1, -1)$ satisfy the constraint.

Thus maximum subject to constraint occurs at $(1, 1)$ and $(-1, -1)$.

$$\text{The maximum is } f(x, y) = -(\pm 1)^2 - (\pm 1)^2 + 25 = 23$$

minimize $-x^2 - y^2 + 25$ on $xy = 1$

Contour plot:



Applications:

Find the points on the surface of $z = \frac{2}{xy^2}$ that are closest to the origin.

Minimize $d^2 = f(x, y, z) = (x - 0)^2 + (y - 0)^2 + (z - 0)^2$ subject to $g(x, y, z) = zxy^2 - 2 = 0$

$$2x = cz y^2 \text{ and } 2y = c(2xy z) \text{ and } 2z = c(xy^2)$$

$$2x2x = 2cxzy^2 \text{ and } 2yy = c(2xyyz) \text{ and } 2z2z = 2c(zxy^2)$$

$$4x^2 = 2y^2 = 4z^2 \text{ implies } 2x^2 = y^2 = 2z^2$$

$$zxy^2 = 2$$

$$z(\pm z)(2z^2) = 2. \text{ Thus } x = z = 1, -1, y = \pm\sqrt{2}$$

$$(x, y, z) = (1, \sqrt{2}, 1), (-1, \sqrt{2}, -1), (1, -\sqrt{2}, 1), (-1, -\sqrt{2}, -1)$$

12.10: 2nd order derivative test

Suppose $z = f(x, y)$

Recall the derivative matrix of f is $Df = \begin{bmatrix} \frac{\partial f}{\partial x} & \frac{\partial f}{\partial y} \end{bmatrix}$

Hessian matrix =

$$D^2f = \begin{bmatrix} \frac{\partial}{\partial x}(\frac{\partial f}{\partial x}) & \frac{\partial}{\partial y}(\frac{\partial f}{\partial x}) \\ \frac{\partial}{\partial y}(\frac{\partial f}{\partial x}) & \frac{\partial}{\partial y}(\frac{\partial f}{\partial y}) \end{bmatrix} = \begin{bmatrix} \frac{\partial^2 f}{\partial x^2} & \frac{\partial^2 f}{\partial y \partial x} \\ \frac{\partial^2 f}{\partial x \partial y} & \frac{\partial^2 f}{\partial y^2} \end{bmatrix} = \begin{bmatrix} \frac{\partial^2 f}{\partial x^2} & \frac{\partial^2 f}{\partial x \partial y} \\ \frac{\partial^2 f}{\partial y \partial x} & \frac{\partial^2 f}{\partial y^2} \end{bmatrix} = Hf$$

Determinant of the Hessian = $\frac{\partial^2 f}{\partial x^2} \frac{\partial^2 f}{\partial y^2} - \left[\frac{\partial^2 f}{\partial y \partial x} \right]^2 = f_{xx}f_{yy} - [f_{xy}]^2$

Recall $f_{xy} = (f_x)_y = (f_y)_x = f_{yx}$ if 2nd order partials are continuous

Theorem 1: **Two-variable second derivative test**

If 2nd order partials of f are continuous in a neighborhood of a critical point (a, b) , then

1. If $\det(Hf(a, b)) = \Delta > 0$, then
 - if $f_{xx} > 0$, then $f(a, b)$ is a local minimum.
 - if $f_{xx} < 0$, then $f(a, b)$ is a local maximum.
2. If $\det(Hf(a, b)) = \Delta < 0$, then $f(a, b)$ is neither a local minimum nor a local maximum

Note if $\det Hf(a, b) = \Delta = 0$, then the 2nd derivative test gives no information.

Example: $f(x, y) = f(x, y) = x^2 + y^2 + pxy$

See chalkboard/<https://www.khanacademy.org/math/multivariable-calculus/application-of-multivariable-derivatives/optimizing-multivariable-functions/a/second-partial-derivative-test>