

Understanding the Hessian.

We now know that

$$f(\vec{x}_0 + \vec{v}) \approx f(x_0) + \langle \vec{v}, \nabla f(x_0) \rangle + \frac{1}{2} \langle \vec{v}, Hf(x_0) \vec{v} \rangle$$

The gradient vector is easy to interpret:
it points "straight uphill" and its norm
is the slope of the hill.

What does $Hf(x_0)$ tell us? We'll need
some more linear algebra.

Definition. If A is an $n \times n$ matrix,
then $\vec{v} \in \mathbb{R}^n$ is called an ~~eigenvector~~
eigenvector of A with eigenvalue λ
if

$$A\vec{v} = \lambda\vec{v}.$$

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Example.

$$A = \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix}$$

eigenvectors

$$e_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, e_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

eigen values

$$\lambda_1, \lambda_2$$

Proposition. The eigenvalues of $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$
 are the roots of the polynomial

$$\det(A - \lambda I) = (a - \lambda)(d - \lambda) - bc.$$

Proof. λ is an eigenvalue of $A \Leftrightarrow$
 there is some \vec{v} with $A\vec{v} = \lambda\vec{v}$ or
 $(A - \lambda I)\vec{v} = 0$. The matrix $(A - \lambda I)$ has
 zero determinant \Leftrightarrow there is some such \vec{v} . \square

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Here is a useful fact:

$\frac{1}{2}$ Spectral Theorem. If A is a symmetric, real $n \times n$ matrix then \exists an orthonormal basis $\vec{v}_1, \dots, \vec{v}_n$ for \mathbb{R}^n so that $A\vec{v}_i = \lambda_i \vec{v}_i$ for all $i \in 1, \dots, n$. All \vec{v}_i and λ_i are real.

This is often summarized as "every symmetric matrix can be diagonalized."

Homework. Suppose $f: \mathbb{R}^n \rightarrow \mathbb{R}$ and $B: \mathbb{R}^n \rightarrow \mathbb{R}^n$ is a linear map^a. Then

$$\nabla(f(B(\vec{x}))) = B \nabla f$$

and

$$H(f(B\vec{x})) = B^T Hf B$$

^a Equivalently, B is an $n \times n$ matrix.

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Now let's return to Taylor's theorem.

Proposition. If $f: \mathbb{R}^n \rightarrow \mathbb{R}$ and $\vec{x}_0 \in \mathbb{R}^n$
 there is an orthonormal basis for \mathbb{R}^n
 so that (in these coordinates) $Hf(\vec{x}_0)$
 is a diagonal matrix.

Proof. Combine homework with Spectral
 theorem.

Proposition. If $f: \mathbb{R}^n \rightarrow \mathbb{R}$ and $\vec{x}_0 \in \mathbb{R}^n$
 there is an orthonormal basis for \mathbb{R}^n
 so that

$$f(\vec{x}_0 + \vec{v}) \approx f(\vec{x}_0) + \langle \nabla f(\vec{x}_0), \vec{v} \rangle + \frac{1}{2} \langle \vec{v}, Hf(\vec{x}_0) \vec{v} \rangle$$

~~as per~~

$$\approx c + b_1 v_1 + \dots + b_n v_n + a_1 v_1^2 + \dots + a_n v_n^2$$

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Here the a_i are the eigenvalues of $Hf(\vec{x}_0)$ (which are the same in any orthonormal basis for \mathbb{R}^n).

Definition. A surface in the form

$z = ax^2 + 2bxy + cy^2$ is called a paraboloid.

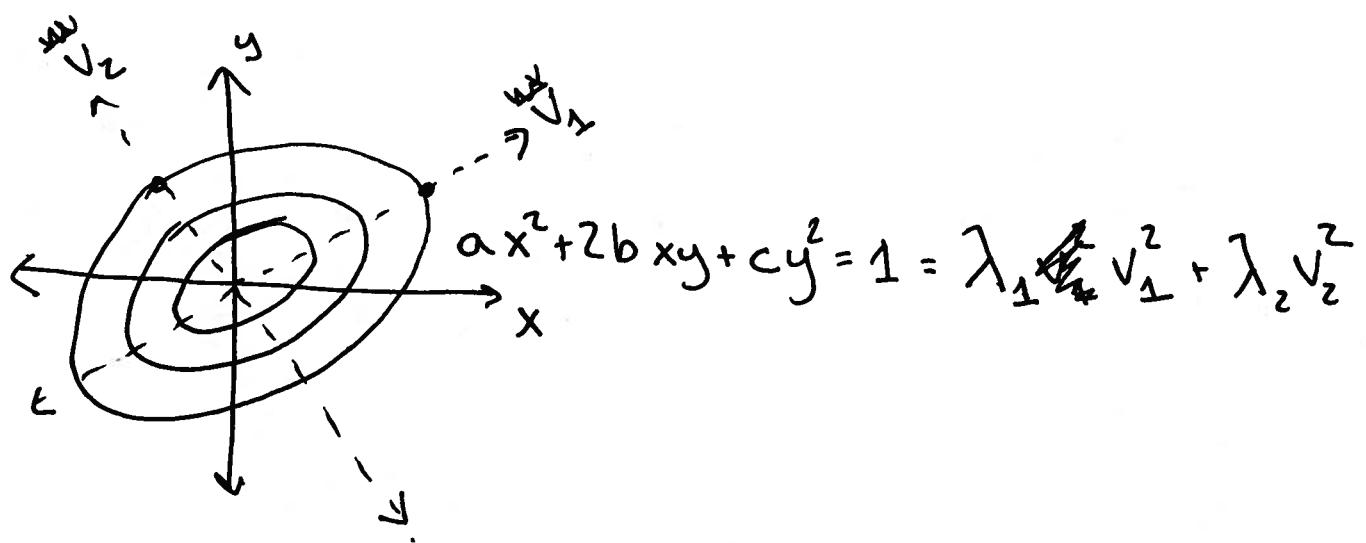
Definition. If the eigenvalues λ_1, λ_2 of $\begin{bmatrix} a & b \\ b & c \end{bmatrix}$ have the same sign^a, the surface is called an elliptic paraboloid.

The level curves of an elliptic paraboloid are ellipses, with major and minor axes in the directions of the eigenvectors

^a Equivalently, $\det \begin{bmatrix} a & b \\ b & c \end{bmatrix} > 0$.

\vec{v}_1 and \vec{v}_2 of $\begin{bmatrix} a & b \\ b & c \end{bmatrix}$.

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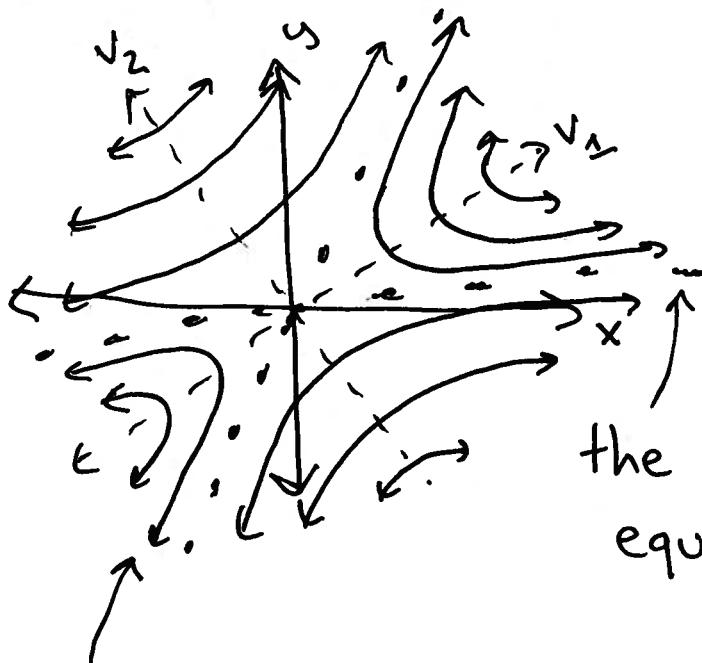


Definition. If the eigenvalues λ_1, λ_2 of $\begin{bmatrix} a & b \\ b & c \end{bmatrix}$ have opposite signs^a, the surface is called a hyperbolic paraboloid.

The level curves of a hyperbolic paraboloid are hyperbolas. The hyperbolas are perpendicular to the eigenvectors \vec{v}_1 and \vec{v}_2 of $\begin{bmatrix} a & b \\ b & c \end{bmatrix}$.

^a Equivalently, $\det \begin{bmatrix} a & b \\ b & c \end{bmatrix} < 0$.

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$$ax^2 + 2bxy + cy^2 = 1 = \lambda_1 V_1^2 + \lambda_2 V_2^2$$

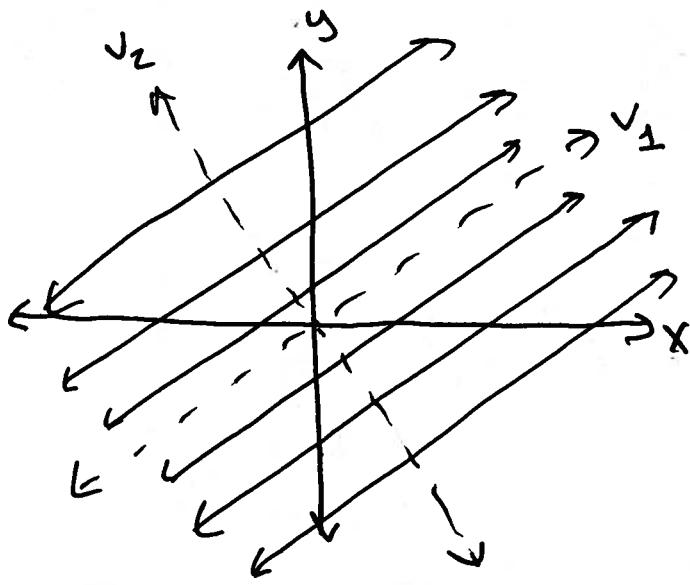
the asymptotic lines have
equation $V_2 = \pm \sqrt{\frac{\lambda_1}{\lambda_2}} V_1$.

the individual hyperbolae
are orthogonal to v_1 and v_2
axes where they cross

Definition. If exactly one of the eigenvalues λ_1, λ_2 of $\begin{bmatrix} a & b \\ b & c \end{bmatrix}$ ~~is~~ zero, the surface is called a cylindrical paraboloid.

The level curves of a cylindrical paraboloid are lines parallel to the direction of the eigenvector with 0 eigenvalue.

(8.)



$$ax^2 + bxy + cy^2 = 1 = \lambda_1 v_1^2 + \lambda_2 v_2^2$$

Note: If both eigenvalues of $\begin{bmatrix} a & b \\ b & c \end{bmatrix}$ are zero then $\begin{bmatrix} a & b \\ b & c \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$ and the surface is the x-y plane.

Let's summarize what we've learned.

For scalar functions, $f: \mathbb{R}^n \rightarrow \mathbb{R}$,

if $n=1$, $f(x_0+v) \approx f(x_0) + f'(x_0)v + \frac{1}{2}f''(x_0)v^2$

so $f(x_0+v)$ is approximated by the parabola

$$f(x_0) + f'(x_0)v + \frac{1}{2}f''(x_0)v^2 = p(v)$$

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The vertex of the parabola is found by solving $p'(v) = 0$ (or completing the square).

$$2 \cdot \frac{1}{2} f''(x_0) v + f'(x_0) = p'(v)$$

so

$$p'(v) = 0 \Leftrightarrow v = -\frac{f'(x_0)}{f''(x_0)}$$

So we could write

$$p(v) = \frac{1}{2} f''(x_0) \left(v + \frac{f'(x_0)}{f''(x_0)} \right)^2 + \left(f(x_0) - \frac{f'(x_0)^2}{2f''(x_0)} \right)$$

$$= \alpha(v - \beta)^2 + \gamma$$

if we wanted to.

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$$\text{If } n > 1, \quad f(\vec{x}_0 + \vec{v}) \approx f(\vec{x}_0) + \langle \vec{v}, \nabla f(\vec{x}_0) \rangle + \frac{1}{2} \langle \vec{v}, Hf(\vec{x}_0) \vec{v} \rangle$$

So $f(\vec{x}_0 + \vec{v})$ is approximated by the paraboloid

$$f(\vec{x}_0) + \langle \vec{v}, \nabla f(\vec{x}_0) \rangle + \frac{1}{2} \langle \vec{v}, Hf(\vec{x}_0) \vec{v} \rangle = p(\vec{v})$$

If we write \vec{v} in the basis of eigenvectors of the $n \times n$ symmetric matrix $Hf(\vec{x}_0)$, then (in this basis) $Hf(\vec{x}_0)$ is a diagonal matrix, and

$$p(\vec{v}) = f(\vec{x}_0) + \sum_{i=1}^n v_i \frac{\partial f(\vec{x}_0)}{\partial v_i} + \frac{1}{2} v_i^2 \frac{\partial^2 f}{\partial v_i^2}(\vec{x}_0)$$

So the center of the paraboloid $p(\vec{v})$ is located at

$$\vec{v}_0 = \left(-\frac{\frac{\partial f}{\partial v_1}(\vec{x}_0)}{\frac{\partial^2 f}{\partial v_1^2}(\vec{x}_0)}, \dots, -\frac{\frac{\partial f}{\partial v_n}(\vec{x}_0)}{\frac{\partial^2 f}{\partial v_n^2}(\vec{x}_0)} \right)$$

(in eigenvector coordinates for $Hf(\vec{x}_0)$).