

This Week : Gradients.

(Chapter 4 : Differentiation of
multivariable functions)

A function $f: \mathbb{R}^n \rightarrow \mathbb{R}$ is called
a "scalar field". To each point

$P = (x_1, x_2, \dots, x_n)$ in \mathbb{R}^n it associates
a scalar

$f(P)$ or $f(x_1, x_2, \dots, x_n)$.

Think: $f(P)$ is temperature at
the point P .

The derivative of f defined as

$$\nabla f = \left\langle \frac{\partial f}{\partial x_1}, \frac{\partial f}{\partial x_2}, \dots, \frac{\partial f}{\partial x_n} \right\rangle$$

$$\left(\text{or } \left\langle \frac{\partial f}{\partial x_1}, \frac{\partial f}{\partial x_2}, \dots, \frac{\partial f}{\partial x_n} \right\rangle \right)$$

Notation : $\frac{\partial f}{\partial x_i}$ means we take

deriv. with respect to variable x_i ,
pretend that the other vars. are constant.

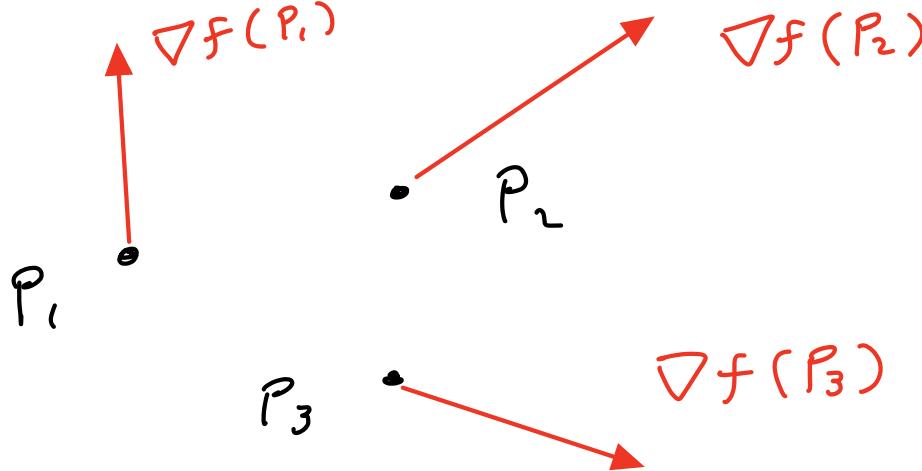
c.g. $f(x, y) = xy$.

$$\nabla f = \left\langle \frac{\partial f}{\partial x}, \frac{\partial f}{\partial y} \right\rangle$$

$$\nabla f(x, y) = \langle y, x \rangle$$

To each point P in \mathbb{R}^n the derivative ∇f associates a vector.

Picture:



Can think of $\nabla f: \mathbb{R}^n \rightarrow \mathbb{R}^n$
as a "vector field".

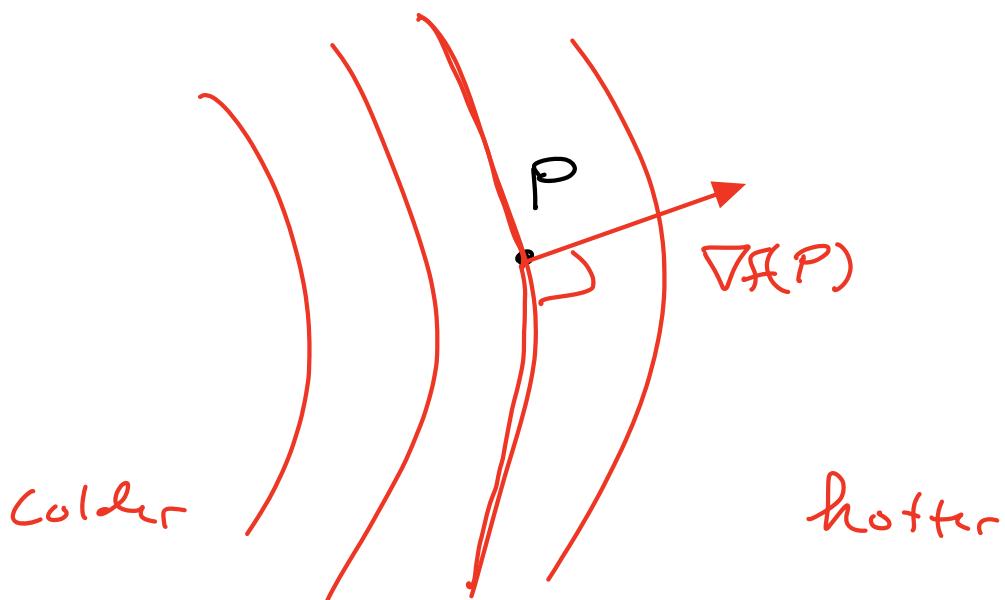
So

" ∇ (scalar field) = vector field."



Meaning: At a point P in \mathbb{R}^n , the vector $\nabla f(P)$ tells us the direction where f is increasing fastest.

e.g. f is temperature



curves of constant temp.

$\nabla f(P)$ points in the direction of increasing temperature & is \perp to the "level curve" at P .

e.g. $f(x,y) = xy$. defined at any point

$$\nabla f(x, y) = \langle y, x \rangle.$$

Pick this point

Consider point $\textcircled{P = (3, 1)}$.

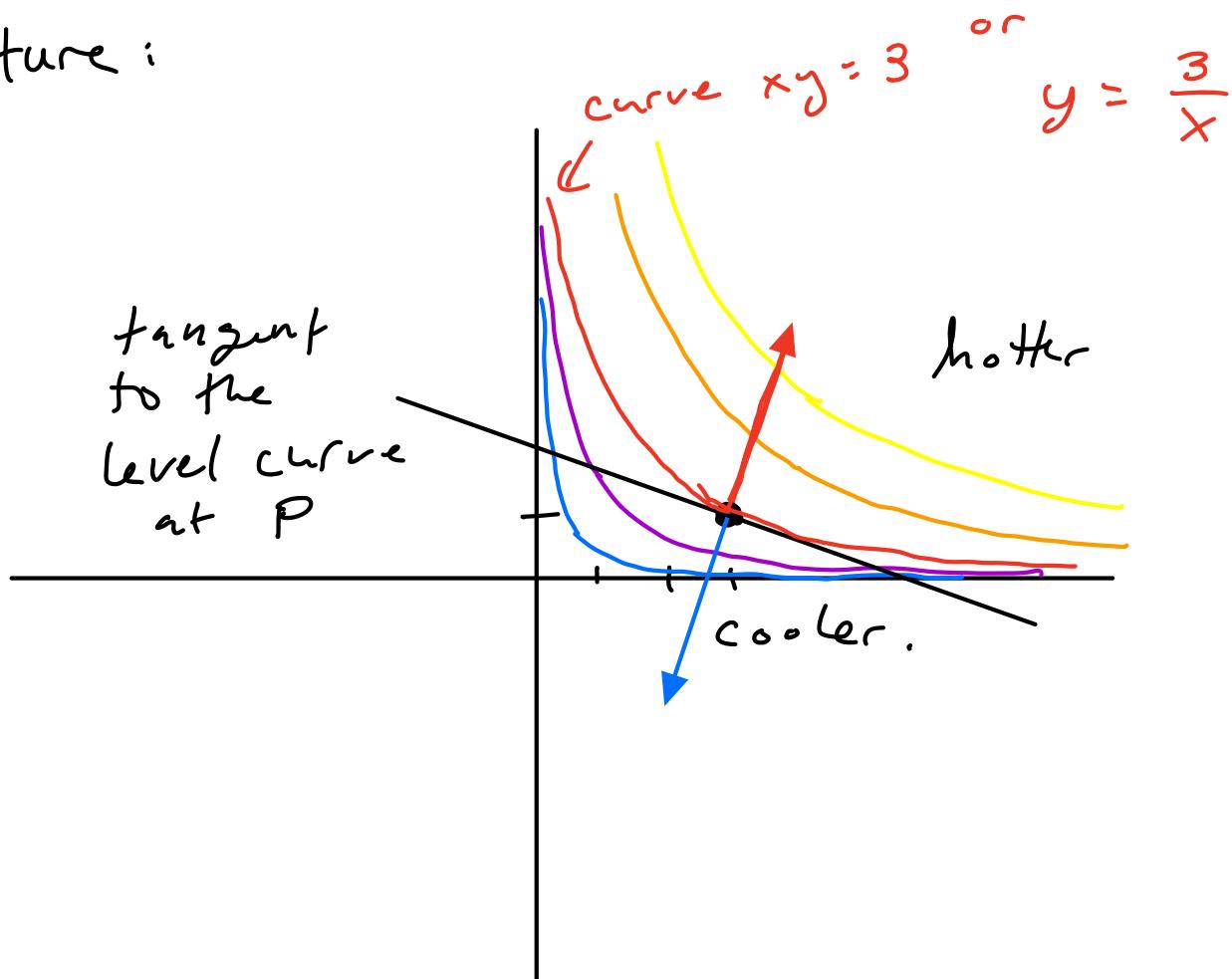
$$f(P) = 3 \cdot 1 = 3.$$

"Temperature at P is 3".

Temperature increases fastest
in the direction $\nabla f(P) =$

$$\nabla f(3, 1) = \langle 1, 3 \rangle$$

Picture:



- $\nabla f(P)$ is the direction in which

f decreases most rapidly at P .



The "same story" in 3D.

Say $f(x, y, z) = 5x^2 - 3xy + xyz$

is temperature at point $P = (x, y, z)$.

$$\nabla f = \left\langle \frac{\partial f}{\partial x}, \frac{\partial f}{\partial y}, \frac{\partial f}{\partial z} \right\rangle$$

$$\frac{\partial f}{\partial x} = 10x - 3y + yz$$

$$\frac{\partial f}{\partial y} = 0 - 3x + xz$$

$$\frac{\partial f}{\partial z} = 0 - 0 + xy$$

$$\nabla f(x, y, z)$$

$$= \langle 10x - 3y + yz, -3x + xz, xy \rangle.$$

e.g. Temperature at $P = (1, 1, 1)$ is

$$\begin{aligned}f(1, 1, 1) &= 5(1)^2 - 3(1)(1) + (1)(1)(1) \\&= 5 - 3 + 1 = 3\end{aligned}$$

and increases most rapidly in
the direction

$$\nabla f(1,1,1) = \langle 10 - 3 + 1, -2 + 1, 1 \rangle \\ = \langle 8, -2, 1 \rangle.$$

Why does it work?

KEY: Multivariable Chain Rule.

Recall the good old chain rule.

$$f: \mathbb{R} \rightarrow \mathbb{R}$$

$$g: \mathbb{R} \rightarrow \mathbb{R}.$$

$$(f \circ g)(t) = f(g(t)).$$

$$(f \circ g)'(t) = f'(g(t)) \cdot g'(t)$$

$$= (f' \circ g)(t) \cdot g'(t)$$

In Chapter 3 we had a new version.

$$\vec{r}: \mathbb{R} \rightarrow \mathbb{R}^n$$

$$f: \mathbb{R} \rightarrow \mathbb{R}$$

$$(\vec{r} \circ f)(t) = \vec{r}(f(t)).$$

$$(\vec{r} \circ f)'(t) = \underbrace{\vec{r}'(f(t))}_{\text{vector}} \cdot \underbrace{f'(t)}_{\text{scalar}}.$$

Now we have something new.

$$f: \mathbb{R}^n \rightarrow \mathbb{R}$$

$$\vec{r}: \mathbb{R} \rightarrow \mathbb{R}^n$$

Say $\vec{r}(t)$ is a parameterized path
in \mathbb{R}^n , say f is temperature.

So $f(\vec{r}(t))$ is the temperature
we feel at time t . We can
think of this as a composition:

$$(f \circ \vec{r})(t) = f(\vec{r}(t))$$

$$f \circ \vec{r}: \mathbb{R} \rightarrow \mathbb{R}$$

$t \mapsto$ our temp. at time t

as we travel the curve.

Question: What temperature change
do we feel?

$$(f \circ \vec{r})'(t) = ?$$

Theorem:

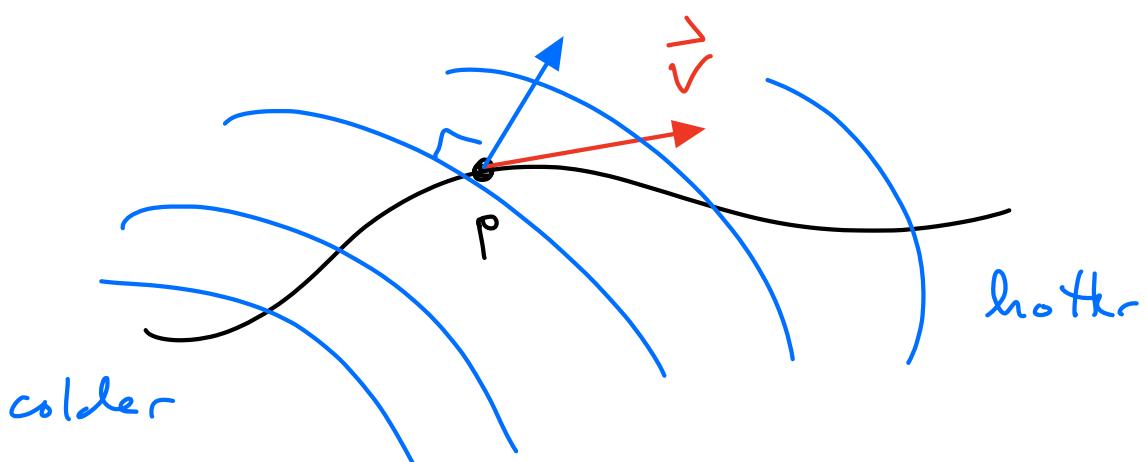
dot product.

$$(f \circ \vec{r})'(t) = \underbrace{\nabla f(\vec{r}(t))}_{\text{vector}} \cdot \underbrace{\vec{r}'(t)}_{\text{vector}}$$

e.g. The rate of change of f
at the point P in the direction
of vector \vec{v} . Suppose $\vec{r}(0) = P$

$$\nabla f(P)$$

$$\vec{r}'(0) = \vec{v}$$



Our rate of change of temp at time $t=0$ is

$$(f \circ \vec{r})'(0)$$

$$= \nabla f(\vec{r}(0)) \cdot \vec{r}'(0)$$

$$= \nabla f(P) \cdot \vec{v}.$$

$$= \underbrace{(\text{gradient})}_{\text{maximized}} \circ (\text{velocity}).$$

maximized when our velocity points in direction of gradient.



Another consequence: The gradient vector at P is \perp to the level set of f at P .

e.g. Consider $f(x, y, z) = (2x)^2 + y^2 + z^2$.

Temperature at point $P = (2, 3, 5)$ is

$$f(2, 3, 5) = (2 \cdot 2)^2 + 3^2 + 5^2 = 50.$$

Gradient vector is

$$\nabla f(x, y, z) = \langle 8x, 2y, 2z \rangle.$$

$$\nabla f(2, 3, 5) = \langle 16, 6, 10 \rangle.$$

The level set $f(x, y, z) = 50$

is the set of points where

temperature = 50.

$$(2x)^2 + y^2 + z^2 = 50.$$

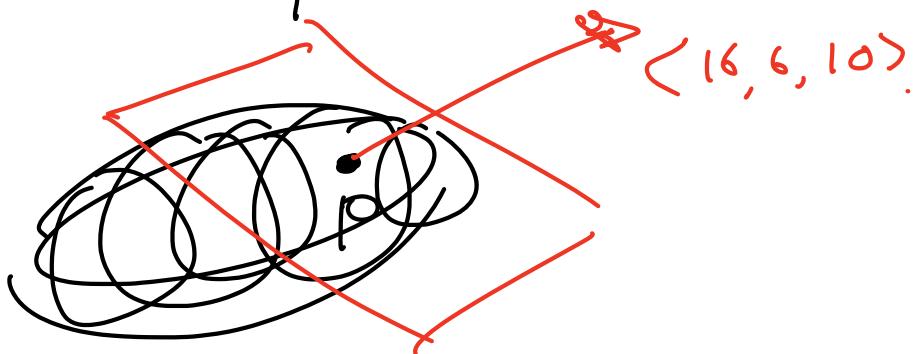
What kind of shape?

Fixed $x \rightarrow$ circle in y, z .

Fixed $y \rightarrow$ ellipse in x, z

Fixed $z \rightarrow$ ellipse in x, y .

Called an "ellipsoid"



Theorem: $\langle 16, 6, 10 \rangle$ is normal
to the tangent plane to the
level surface at $P = (2, 3, 5)$.

Conclusion:

Equation of tangent plane to
ellipsoid $(2x)^2 + y^2 + z^2 = 50$
at the point $(2, 3, 5)$ is

$$\nabla f(2, 3, 5) \cdot \langle x-2, y-3, z-5 \rangle = 0.$$

$$\langle 16, 6, 10 \rangle \cdot \langle x-2, y-3, z-5 \rangle = 0$$

$$16(x-2) + 6(y-3) + 10(z-5) = 0$$

:

$$8x + 3y + 5z = 50.$$