§12.5: Chain Rule

Recall the chain rule for single-variable functions:

$$\frac{d}{dt}f(x(t)) = f'(x(t))x'(t)$$
, i.e. $\frac{df}{dt} = \frac{df}{dx}\frac{dx}{dt}$.

Here's an informal way to understand the chain rule.

The linearisation of f says:

$$f(x + \delta x) \approx f(x) + f'(x)\delta x.$$
 (*)

Write $x + \delta x$ for $x(t + \delta t)$. Using the linearisation of x:

$$x + \delta x = x(t + \delta t) \approx x(t) + x'(t)\delta t$$

 $\delta x \approx x'(t)\delta t$

Substituting into (*):

$$f(x(t+\delta t)) \approx f(x(t)) + f'(x(t))x'(t)\delta t.$$

Compare the above to the linearisation of the composite function f(x(t)):

$$f(x(t+\delta t)) \approx f(x(t)) + \left| \frac{d}{dt} f(x(t)) \right| \delta t.$$

So the quantities in the blue rectangles should be the same.

Now we derive a simple example of a multivariate chain rule in the same way.

Imagine that you are walking on \mathbb{R}^2 , and your position at time t is (x(t),y(t)). The temperature at the point (x,y) is f(x,y). So the temperature that you feel at time t is the composite function f(x(t),y(t)). What is $\frac{d}{dt}f((x(t),y(t)))$, the rate of change of temperature that you feel?

The linearisation of the temperature function is

$$f(x + \delta x, y + \delta y) \approx f(x, y) + \frac{\partial f}{\partial x} \delta x + \frac{\partial f}{\partial y} \delta y.$$

And the linearisations of x and y tell us that

$$\delta x \approx \frac{dx}{dt} \delta t; \quad \delta y \approx \frac{dy}{dt} \delta t.$$

Substituting into (*)

$$f(x(t+\delta t), y(t+\delta t)) \approx f(x,y) + \frac{\partial f}{\partial x} \frac{dx}{dt} \delta t + \frac{\partial f}{\partial y} \frac{dy}{dt} \delta t.$$

Comparing with the linearisation of f(x(t), y(t)):

$$\frac{df}{dt} = \frac{\partial f}{\partial x} \frac{dx}{dt} + \frac{\partial f}{\partial y} \frac{dy}{dt}$$

(This is not a rigorous proof because we haven't checked that the errors are small enough. We sketch a rigorous and more general version of this argument on p10. For a different rigorous proof, see the first page of §12.5 in the textbook.)

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Example: Let $f(x,y) = xy^2$, and $x = \ln t, y = 3t^2$.

Find $\frac{df}{dt}$.

$$\frac{df}{dt} = \frac{\partial f}{\partial x} \frac{dx}{dt} + \frac{\partial f}{\partial y} \frac{dy}{dt}.$$

We showed that, if f(x,y) is a 2-variable function, and x and y are functions of t, then

$$\frac{df}{dt} = \frac{\partial f}{\partial x}\frac{dx}{dt} + \frac{\partial f}{\partial y}\frac{dy}{dt}.$$

Now suppose x, y are multivariate functions, e.g. x(s, t), y(s, t).

To find $\frac{\partial f}{\partial t}$, we treat s as a constant throughout, so

$$\frac{\partial f}{\partial t} = \frac{\partial f}{\partial x} \frac{\partial x}{\partial t} + \frac{\partial f}{\partial y} \frac{\partial y}{\partial t};$$

And similarly:

$$\frac{\partial f}{\partial s} = \frac{\partial f}{\partial x} \frac{\partial x}{\partial s} + \frac{\partial f}{\partial y} \frac{\partial y}{\partial s}.$$

Example: Let $f(x,y) = xy^2$, and $x(s,t) = \ln(s+t)$, $y(s,t) = 3t^2\cos s$. Find $\frac{\partial f}{\partial s}$ and $\frac{\partial f}{\partial s}(0,1)$. $\frac{\partial f}{\partial s} = \frac{\partial f}{\partial x}\frac{\partial x}{\partial s} + \frac{\partial f}{\partial y}\frac{\partial y}{\partial s}.$

$$\frac{\partial f}{\partial s} = \frac{\partial f}{\partial x} \frac{\partial x}{\partial s} + \frac{\partial f}{\partial y} \frac{\partial y}{\partial s}$$

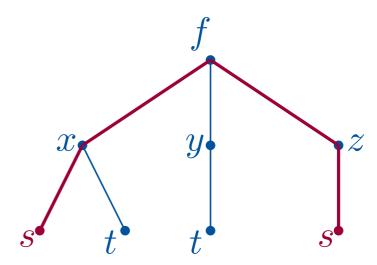
In ex. sheet #16 Q2, we are given f(x,y,z) and $x(s,t)=e^{st}$, $y(s,t)=t^2$, $z(s,t)=s^2+1$. The chain rule says

$$\frac{\partial f}{\partial s} = \frac{\partial f}{\partial x} \frac{\partial x}{\partial s} + \frac{\partial f}{\partial y} \frac{\partial y}{\partial s} + \frac{\partial f}{\partial z} \frac{\partial z}{\partial s},$$

but the second term (in y) is unnecessary because y does not depend on s. To simplify things in such cases, we can draw a dependency chart showing which functions depend on which variables. Then the terms in the chain rule for $\frac{\partial f}{\partial s}$

correspond to all the paths from s to f.

Dependency charts can be really useful when there are many variables, or when dealing with a triple composition (e.g. if s and t here are functions of u, v, w).



We can compute higher order derivatives of composite functions by applying the chain rule repeatedly.

Example: Let f(x,y) be a two variable function, and x=2s+3t, y=st. Find an expression for $\frac{\partial^2}{\partial s \partial t} f(x(s,t),y(s,t))$ in terms of the partial derivatives of f.

The chain rule in terms of Jacobian matrices and the derivative linear transformation

Remember from p4 that, for f(x,y), x(s,t), y(s,t), we have

$$\frac{\partial f}{\partial s} = \frac{\partial f}{\partial x} \frac{\partial x}{\partial s} + \frac{\partial f}{\partial y} \frac{\partial y}{\partial s}; \quad \frac{\partial f}{\partial t} = \frac{\partial f}{\partial x} \frac{\partial x}{\partial t} + \frac{\partial f}{\partial y} \frac{\partial y}{\partial t}.$$

In the notation of Jacobian matrices, we have

$$Df(s,t) = \begin{pmatrix} \frac{\partial f}{\partial s} & \frac{\partial f}{\partial t} \end{pmatrix} = \begin{pmatrix} \frac{\partial f}{\partial x} & \frac{\partial f}{\partial y} \end{pmatrix} \begin{pmatrix} \frac{\partial x}{\partial s} & \frac{\partial x}{\partial t} \\ \frac{\partial y}{\partial s} & \frac{\partial y}{\partial t} \end{pmatrix} = Df(x(s,t), y(s,t))D\mathbf{g}(s,t),$$

writing g(s,t) for (x(s,t),y(s,t)) (i.e. $g_1=x$ and $g_2=y$).

In general, the Jacobian matrix of a composite function is the matrix product of the Jacobian matrices $D(\mathbf{f} \circ \mathbf{g})(\mathbf{t}) = D\mathbf{f}(\mathbf{g}(\mathbf{t}))D\mathbf{g}(\mathbf{t}).$

Because the product of matrices correspond to the composition of linear transformations, this says that the derivative of a composition is a composition of the derivatives.

Example: Let $\mathbf{g}: \mathbb{R}^2 \to \mathbb{R}^3$ be a function such that

$$\mathbf{g}(1,2) = (1,2,1) \text{ and } D\mathbf{g}(1,2) = \begin{pmatrix} 1/2 & 1/2 \\ 4 & 1 \\ 1 & 0 \end{pmatrix}.$$

Let $\mathbf{f}: \mathbb{R}^3 \to \mathbb{R}^2$ be given by $\mathbf{f}(x, y, z) = (x^2 e^y, y^2 z)$. Find $D(\mathbf{f} \circ \mathbf{g})(1, 2)$.

 $D(\mathbf{f} \circ \mathbf{g})(\mathbf{t}) = D\mathbf{f}(\mathbf{g}(\mathbf{t}))D\mathbf{g}(\mathbf{t}).$

Non-examinable: the proof of the chain rule (different from the textbook)

The main idea is the linearisation argument on pp1-2. We will show carefully that the errors in the linearisation are small compared to $|\delta \mathbf{t}|$, as required in the definition of the derivative.

We wish to show that $D(\mathbf{f} \circ \mathbf{g})(\mathbf{t}) = D\mathbf{f}(\mathbf{g}(\mathbf{t}))D\mathbf{g}(\mathbf{t})$. So we need to show that $D\mathbf{f}(\mathbf{g}(\mathbf{t}))D\mathbf{g}(\mathbf{t})$ satisfies the definition of the derivative $D(\mathbf{f} \circ \mathbf{g})$, i.e.

$$\frac{(\mathbf{f} \circ \mathbf{g})(\mathbf{t} + \delta \mathbf{t}) - (\mathbf{f} \circ \mathbf{g})(\mathbf{t}) - [D\mathbf{f}(\mathbf{g}(\mathbf{t}))][D\mathbf{g}(\mathbf{t})]\delta \mathbf{t}}{|\delta \mathbf{t}|} \to 0 \text{ as } \delta \mathbf{t} \to \mathbf{0}.$$

Let $\mathbf{x} = \mathbf{g}(\mathbf{t})$ and $\mathbf{x} + \delta \mathbf{x} = \mathbf{g}(\mathbf{t} + \delta \mathbf{t})$, and rewrite the expression above as

$$\frac{\mathbf{f}(\mathbf{g}(\mathbf{t} + \delta \mathbf{t})) - \mathbf{f}(\mathbf{g}(\mathbf{t})) - [D\mathbf{f}(\mathbf{g}(\mathbf{t}))]\delta \mathbf{x}}{|\delta \mathbf{t}|} + \frac{[D\mathbf{f}(\mathbf{g}(\mathbf{t}))]\delta \mathbf{x} - [D\mathbf{f}(\mathbf{g}(\mathbf{t}))][D\mathbf{g}(\mathbf{t})]\delta \mathbf{t}}{|\delta \mathbf{t}|}$$

$$= \frac{\mathbf{f}(\mathbf{x} + \delta \mathbf{x}) - \mathbf{f}(\mathbf{x}) - [D\mathbf{f}(\mathbf{x})]\delta \mathbf{x}}{|\delta \mathbf{x}|} \frac{|\delta \mathbf{x}|}{|\delta \mathbf{t}|} + [D\mathbf{f}(\mathbf{g}(\mathbf{t}))] \left(\frac{\delta \mathbf{x} - [D\mathbf{g}(\mathbf{t})]\delta \mathbf{t}}{|\delta \mathbf{t}|} \right)$$

goes to 0 because $D\mathbf{f}$ $_{HKBU\ Ma}$ is the derivative of f.

is finite because x = gis differentiable.

goes to 0 because $D\mathbf{g}$ is the derivative of g = x.