

# Lecture 4 - Calculus II

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# Morning challenge!

Take the derivative of the following functions (it is not necessary to simplify):

▶  $f(x) = x^3 + 6x^2 + 3$

▶  $f(x) = \frac{x^3+5x}{x^2-2}$

▶  $f(x) = \ln(3x^4 + \sqrt[5]{x^4 - 3x^2})$

▶  $f(x) = e^{e^{5x^3+4x}}$

▶  $f(x) = ax^2 + bx + c$ , with  $a$ ,  $b$  and  $c$  constants.

# Agenda

Integration

Multivariate calculus

# Agenda

Integration

Multivariate calculus

What is the area under the curve?

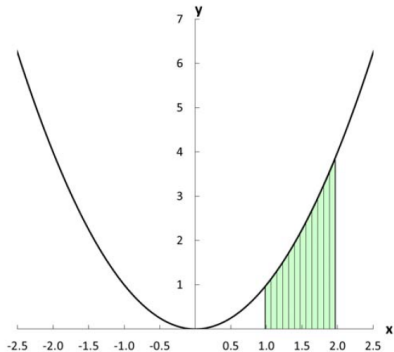
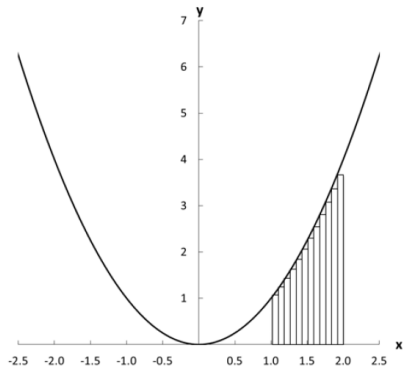


Figure 7.1: Area under  $y = x^2$  from  $x = 1$  to  $x = 2$



# Integrals

**Definite Integral:** An operator that helps us find the area under a curve.

$$\int_a^b f(x)dx$$

What area? The area of the region bounded by  $f(x)$  and the  $x$  axis on the domain  $a$  to  $b$ .

The  $f(x)$ , or more generally, the expression multiplying the  $dx$ , is known as the *integrand*. The  $dx$  tells you the variable of integration.

When unbounded, it is called **indefinite integral** and represents the antiderivative:

- ▶ The definite integral returns a value, the area under the curve.
- ▶ The indefinite integral returns a function that, when differentiated, reproduces the integrand.

# Fundamental theorem of calculus

$$\int_a^b f(x)dx = F(b) - F(a)$$

Area under the curve = Antiderivative (Upper bound) - Antiderivative (Lower bound)

# Antiderivative

The antiderivative is the function whose derivative is  $f(x)$ .

$$F(x) = \int f(x)dx$$

This is the same as an “indefinite integral”.

Differentiation and antidifferentiation are inverse operations of one another.

$$\frac{dF(x)}{dx} = f(x)$$



# Do not forget the C!

- ▶ When you take the derivative of a constant, it is zero.
- ▶ There is not enough information in a derivative for us to reverse engineer what that constant was in the original function.
- ▶ SO whenever you take an antiderivative, you must include  $+C$  at the end to note that there may be a constant.

Example: Suppose  $f(x) = 1$ . What is its antiderivative?

Following the rules of differentiation, we know that  $F(x)$  at least contains  $x$ , because  $\frac{dx}{dx} = 1$ .

However, we do not know anything about the constant. It could be  $F(x) = x + 1$  or  $F(x) = x + 100$ . And due to we do not know, we say:

$$F(x) = x + c$$

# More simple examples

**Example 1:** Suppose  $f(x) = x^3$ . What is its antiderivative?

Following the rules of differentiation, we know that  $\frac{dx^4}{dx} = 4x^3$  and this is similar to the  $x^3$  that we are looking for.

So if we cancel the 4, we will get our result.

Therefore:  $F(x) = \frac{x^4}{4} + c$  is the antiderivative!

Check your answer:  $\frac{dF(x)}{dx} = \frac{\cancel{4}x^3}{\cancel{4}} = x^3 \checkmark \checkmark$

# Rules of integration

Table 7.1: List of Rules of Integration

Fundamental theorem of calculus	$\int_a^b f(x)dx = F(b) - F(a)$
Rules for bounds	$\int_a^b f(x)dx = -\int_b^a f(x)dx$ $\int_a^a f(x)dx = 0$ $\int_a^b f(x)dx = \int_a^c f(x)dx + \int_c^b f(x)dx$ for $c \in [a, b]$
Linear rule	$\int (af(x) + bg(x))dx = a \int f(x)dx + b \int g(x)dx$
Integration by substitution	$\int_a^b f(g(u))g'(u)du = \int_{g(a)}^{g(b)} f(x)dx$
Integration by parts	$\int f(x)g'(x)dx = f(x)g(x) - \int f'(x)g(x)dx$
Power rule 1	$\int x^n dx = \frac{x^{n+1}}{n+1} + C$ if $n \neq -1$
Power rule 2	$\int x^{-1} dx = \ln x  + C$
Exponential rule 1	$\int e^x dx = e^x + C$
Exponential rule 2	$\int a^x dx = \frac{a^x}{\ln(a)} + C$
Logarithm rule 1	$\int \ln(x)dx = x \ln(x) - x + C$
Logarithm rule 2	$\int \log_a(x)dx = \frac{x \ln(x) - x}{\ln(a)} + C$
Trigonometric rules	$\int \sin(x)dx = -\cos(x) + C$ $\int \cos(x)dx = \sin(x) + C$ $\int \tan(x)dx = -\ln( \cos(x) ) + C$
Piecewise rules	Split definite integral into corresponding pieces

## More simple examples

**Example 2:** Calculate the area under the curve  $x^2$  from 3 to 9

From the fundamental theorem of calculus we know that the answer is given by:

$$\int_3^9 f(x)dx = F(9) - F(3)$$

The next step is to determinate  $F(x)$ . Using the power rule, we know that  $\int x^n dx = \frac{x^{n+1}}{n+1}$  so the antiderivative is  $\frac{x^3}{3}$ .

Replacing that in our initial expression:

$$\int_3^9 f(x)dx = \frac{9^3}{3} - \frac{3^3}{3}$$

$$\int_3^9 f(x)dx = 243 - 9$$

$$\int_3^9 f(x)dx = 234 \checkmark \checkmark$$

# Integration by substitution

Integration by substitution is attempted whenever the integral contains a composite function that one cannot integrate easily.

In particular, when integrating by substitution, look for whether there is a function and something that looks like it could be the function's derivative.

$$\int_a^b f(g(u))g'(u)du = \int_{g(a)}^{g(b)} f(x)d(x)$$

## Example

Calculate  $\int_1^2 2x(x^2 + 1)^3 dx$

- ▶ Answer: Let's define  $u = x^2 + 1$
- ▶ Then,  $\frac{du}{dx} = 2x$ . Therefore, we could say:  $du = 2x dx$
- ▶ Substituting in our original expression:  $\int_1^2 u^3 du$
- ▶ However, the bounds were defined for  $x$ , not for  $u$ . So we have to adapt them to  $u$ .
  - Lower bound:  $(1)^2 + 1 = 2$
  - Upper bound:  $(2)^2 + 1 = 5$
- ▶ Finally, the right expression to calculate is:  $\int_2^5 u^3 du$
- ▶ Final answer:  $\int_2^5 u^3 du = \frac{5^4}{4} - \frac{2^4}{4} = 152.25$

# Agenda

Integration

Multivariate calculus

# Multivariate calculus

It is just calculus with more than one variable in the function.

$$f(x, y, z) = \frac{xy^2}{zy+x} - \frac{y}{x+4} + 1$$

$$f(K, L) = K^\alpha L^{1-\alpha}$$

$$f(x_1, x_2, x_3) = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_3$$



# Partial derivatives

Treat every variable other than  $x$  as a constant, then take the derivative with respect to  $x$ .

$$f(x, y, z) = 3x^2y + zy^3 + \ln(y) - x$$
$$\frac{df(x, y, z)}{dx} = 6xy - 1$$

You can take a partial derivative for any variable in the function:

$$\frac{df(x, y, z)}{dy} = 3x^2 + 3zy^2 + \frac{1}{y}$$

**Key:** The bottom of the derivative notation tells you which variable you will be taking the derivative with respect to!

# Gradients

A vector of all possible first order derivatives.

For  $f(x,y,z)$  the gradient  $\nabla = \begin{bmatrix} dx \\ dy \\ dz \end{bmatrix}$

Simply take the first order partial derivative with respect to each variable and arrange as a vector!

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

$$\frac{df(x,y,z)}{dx} = 2x + 5y$$

$$\frac{df(x,y,z)}{dy} = 5x$$

$$\frac{df(x,y,z)}{dz} = 3z^2$$

$$\nabla = \begin{bmatrix} 2x + 5y \\ 5x \\ 3z^2 \end{bmatrix}$$

# Mixed partial derivatives

Take the derivative first with respect to one variable, then take the second derivative with respect to another variable.

Indicated by  $\frac{d^2 f}{dx dy}$  or  $\frac{\partial^2 f}{\partial x \partial y}$  or the notation  $f_{xy}$

$$f(x, y, z) = 3x^2y + zy^3 + \ln(y) - x$$

$$\frac{df(x, y, z)}{dx} = 6xy - 1$$

$$\frac{d^2 f(x, y, z)}{dx dy} = 6x$$

# The Hessian matrix

Hessians are used in optimization of multivariate functions. They tell us how a function behaves in multiple dimensions.

This is basically the second derivative when your function exists in multiple dimensions!

$$\begin{bmatrix} \frac{d^2 f(x,y,z)}{dx^2} & \frac{d^2 f(x,y,z)}{dxdy} & \frac{d^2 f(x,y,z)}{dxdz} \\ \frac{d^2 f(x,y,z)}{dydx} & \frac{d^2 f(x,y,z)}{dy^2} & \frac{d^2 f(x,y,z)}{dydz} \\ \frac{d^2 f(x,y,z)}{dzdx} & \frac{d^2 f(x,y,z)}{dzdy} & \frac{d^2 f(x,y,z)}{dz^2} \end{bmatrix}$$

# The Hessian matrix

Hessians are symmetric matrices

$$\begin{bmatrix} \frac{d^2 f(x,y,z)}{dx^2} & \frac{d^2 f(x,y,z)}{dx dy} & \frac{d^2 f(x,y,z)}{dx dz} \\ \frac{d^2 f(x,y,z)}{dy dx} & \frac{d^2 f(x,y,z)}{dy^2} & \frac{d^2 f(x,y,z)}{dy dz} \\ \frac{d^2 f(x,y,z)}{dz dx} & \frac{d^2 f(x,y,z)}{dz dy} & \frac{d^2 f(x,y,z)}{dz^2} \end{bmatrix}$$

And this is because  $\frac{d^2 f(x,y,z)}{dx dy} = \frac{d^2 f(x,y,z)}{dy dx}$

## Group exercise

Using this function:

$$f(x, y, z) = x + y + z + x^2 y^2 z^2$$

Find the gradient ( $\nabla$ ) and the Hessian.

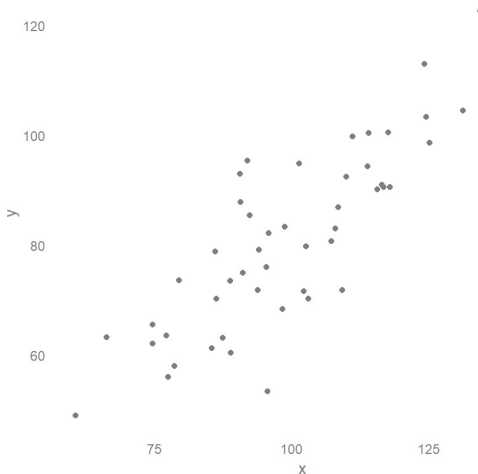
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# Application - Ordinary Least Squares (OLS)

Suppose that you want to understand the following relationship that was generated by the following equation (data generation process):

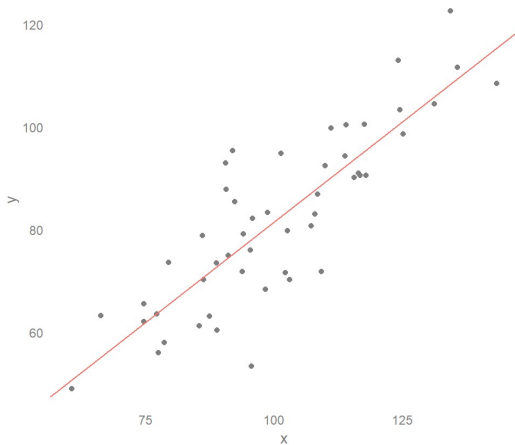
$$Y_i = \alpha + \beta X_i + \varepsilon_i$$





# Application - Ordinary Least Squares (OLS)

OLS help us to identify the linear equation that minimizes the “residuals” squared:  $\hat{Y}_i = \hat{\alpha} + \beta \hat{X}_i$ . What is a residual? The difference between the actual point ( $Y_i$ ) and our estimation of that point ( $\hat{Y}_i$ )



# Application - Ordinary Least Squares (OLS)

What we want is to minimize the residuals:

$$\text{Min}S = \text{Min} \sum_{i=1}^n (e_i^2) = \text{Min} \sum_{i=1}^n ((Y_i - \hat{Y}_i)^2) = \text{Min} \sum_{i=1}^n ((Y_i - \hat{\alpha} - \hat{\beta}X_i)^2)$$

FOC:

$$\frac{\partial S}{\partial \hat{\alpha}} = -2 \sum_{i=1}^n (Y_i - \hat{\alpha} - \hat{\beta}X_i) = 0 \quad (1)$$

$$\frac{\partial S}{\partial \hat{\beta}} = -2 \sum_{i=1}^n X_i (Y_i - \hat{\alpha} - \hat{\beta}X_i) = 0 \quad (2)$$

# Application - Ordinary Least Squares (OLS)

Reordering (1) we have:

$$\begin{aligned} -2 \sum_{i=1}^n (Y_i - \hat{\alpha} - \hat{\beta} X_i) &= 0 \\ \sum_{i=1}^n Y_i &= \hat{\alpha} n + \hat{\beta} \sum_{i=1}^n X_i \\ \hat{\alpha} &= \frac{\sum_{i=1}^n Y_i - \hat{\beta} \sum_{i=1}^n X_i}{n} \\ \hat{\alpha} &= \bar{Y} - \hat{\beta} \bar{X} \end{aligned} \tag{3}$$

# Application - Ordinary Least Squares (OLS)

Reordering (2) and replacing  $\hat{\alpha}$ :

$$\begin{aligned}
 -2 \sum_{i=1}^n X_i (Y_i - \hat{\alpha} - \hat{\beta} X_i) &= 0 \\
 \sum_{i=1}^n X_i Y_i &= \hat{\alpha} \sum_{i=1}^n X_i + \hat{\beta} \sum_{i=1}^n X_i^2 \\
 \sum_{i=1}^n X_i Y_i &= (\bar{Y} - \hat{\beta} \bar{X}) \sum_{i=1}^n X_i + \hat{\beta} \sum_{i=1}^n X_i^2 \\
 \sum_{i=1}^n X_i Y_i - \bar{Y} \sum_{i=1}^n X_i &= -\hat{\beta} \bar{X} \sum_{i=1}^n X_i + \hat{\beta} \sum_{i=1}^n X_i^2 \\
 \sum_{i=1}^n X_i Y_i - \bar{Y} \sum_{i=1}^n X_i &= \hat{\beta} \left( \sum_{i=1}^n X_i^2 - \bar{X} \sum_{i=1}^n X_i \right) \\
 \hat{\beta} &= \frac{\sum_{i=1}^n X_i Y_i - \bar{Y} \sum_{i=1}^n X_i}{\sum_{i=1}^n X_i^2 - \bar{X} \sum_{i=1}^n X_i}
 \end{aligned} \tag{4}$$

# Application - Ordinary Least Squares (OLS)

We had:  $\hat{\beta} = \frac{\sum_{i=1}^n X_i Y_i - \bar{Y} \sum_{i=1}^n X_i}{\sum_{i=1}^n X_i^2 - \bar{X} \sum_{i=1}^n X_i}$

Special property that could be used for  $X$  and  $Y$ :

$$\bar{Y} = \frac{\sum_{i=1}^n Y_i}{n} \quad (5)$$

So, we can rearrange our  $\hat{\beta}$ :

$$\begin{aligned} \hat{\beta} &= \frac{\sum_{i=1}^n X_i Y_i - \frac{\sum_{i=1}^n X_i \sum_{i=1}^n Y_i}{n}}{\sum_{i=1}^n X_i^2 - \frac{\sum_{i=1}^n X_i \sum_{i=1}^n X_i}{n}} \\ \hat{\beta} &= \frac{n \sum_{i=1}^n X_i Y_i - \sum_{i=1}^n X_i \sum_{i=1}^n Y_i}{n \sum_{i=1}^n X_i^2 - (\sum_{i=1}^n X_i)^2} \end{aligned} \quad (6)$$

## Application - Ordinary Least Squares (OLS)

Also, from (5), we know:

$$\sum_{i=1}^n Y_i = \frac{n \sum_{i=1}^n Y_i}{n} = n\bar{Y} \quad (7)$$

And we had:

$$\hat{\beta} = \frac{n \sum_{i=1}^n X_i Y_i - \sum_{i=1}^n X_i \sum_{i=1}^n Y_i}{n \sum_{i=1}^n X_i^2 - (\sum_{i=1}^n X_i)^2}$$

Using (7) on the previous expression:

$$\begin{aligned} \hat{\beta} &= \frac{n \sum_{i=1}^n X_i Y_i - n^2 \bar{X} \bar{Y}}{n \sum_{i=1}^n X_i^2 - n^2 \bar{X}^2} \\ \hat{\beta} &= \frac{n(\sum_{i=1}^n X_i Y_i - n \bar{X} \bar{Y})}{n(\sum_{i=1}^n X_i^2 - n \bar{X}^2)} \\ \hat{\beta} &= \frac{\sum_{i=1}^n X_i Y_i - n \bar{X} \bar{Y}}{\sum_{i=1}^n X_i^2 - n \bar{X}^2} \end{aligned} \quad (8)$$

# Application - Ordinary Least Squares (OLS)

For simplicity, let's define:

$$\phi = \sum_{i=1}^n X_i Y_i - n \overline{X} \overline{Y}$$

$$\phi = \sum_{i=1}^n X_i^2 - n \overline{X}^2.$$

In  $\phi$  we can add and subtract  $n \overline{X} \overline{Y}$ :

$$\phi = \sum_{i=1}^n X_i Y_i - n \overline{X} \overline{Y} - n \overline{X} \overline{Y} + n \overline{X} \overline{Y} \quad (9)$$

Using (7) we have:

$$\phi = \sum_{i=1}^n X_i Y_i - \overline{Y} \sum_{i=1}^n X_i - \overline{X} \sum_{i=1}^n Y_i + \sum_{i=1}^n \overline{X} \overline{Y} \quad (10)$$

# Application - Ordinary Least Squares (OLS)

We had:

$$\phi = \sum_{i=1}^n X_i Y_i - \bar{Y} \sum_{i=1}^n X_i - \bar{X} \sum_{i=1}^n Y_i + \sum_{i=1}^n \bar{X} \bar{Y}$$

Finally, we know that the summation is a linear operator, and therefore:

$$\begin{aligned}\phi &= \sum_{i=1}^n (X_i Y_i - \bar{Y} X_i - \bar{X} Y_i + \bar{X} \bar{Y}) \\ \phi &= \sum_{i=1}^n (X_i (Y_i - \bar{Y}) - \bar{X} (Y_i - \bar{Y})) \\ \phi &= \sum_{i=1}^n (X_i - \bar{X}) (Y_i - \bar{Y})\end{aligned}\tag{11}$$



## Application - Ordinary Least Squares (OLS)

We repeat this procedural for  $\varphi$ , adding and subtracting  $n\bar{X}^2$ :

$$\begin{aligned}\varphi &= \sum_{i=1}^n X_i^2 - n\bar{X}^2 \\ \varphi &= \sum_{i=1}^n X_i^2 - n\bar{X}^2 + n\bar{X}^2 - n\bar{X}^2 \\ \varphi &= \sum_{i=1}^n X_i^2 - \bar{X} \sum_{i=1}^n X_i + n\bar{X}^2 - \bar{X} \sum_{i=1}^n X_i \\ \varphi &= \sum_{i=1}^n X_i^2 - 2\bar{X} \sum_{i=1}^n X_i + n\bar{X}^2 \\ \varphi &= \sum_{i=1}^n X_i^2 - \sum_{i=1}^n 2\bar{X}X_i + \sum_{i=1}^n \bar{X}^2 \\ \varphi &= \sum_{i=1}^n (X_i^2 - 2\bar{X}X_i + \bar{X}^2) \\ \varphi &= \sum_{i=1}^n (X_i - \bar{X})^2\end{aligned}\tag{12}$$

# Application - Ordinary Least Squares (OLS)

Replacing  $\phi$  and  $\varphi$  in (10), we get the traditional form of  $\beta$ :

$$\hat{\beta} = \frac{\sum_{i=1}^n (X_i - \bar{X})(Y_i - \bar{Y})}{\sum_{i=1}^n (X_i - \bar{X})^2} \quad (13)$$