



COMPUTATIONAL FINANCE & RISK MANAGEMENT

UNIVERSITY *of* WASHINGTON

Department of Applied Mathematics

AMATH 460: Mathematical Methods for Quantitative Finance

3. Partial Derivatives

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Outline

- 1 Functions of Several Variables
- 2 Higher Order Partial Derivatives
- 3 Functions of Two Variables
- 4 The Chain Rule for Functions of Several Variables
- 5 Implicit Functions
- 6 Put-Call Parity and The Greeks
- 7 Delta
- 8 Gamma (Γ)
- 9 Rho (ρ) and Vega
- 10 Theta (Θ)

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Scalar Valued Functions

- A function of several variables that takes values in \mathbb{R} is called a scalar valued function.
- Notation: $f : \mathbb{R}^n \rightarrow \mathbb{R}$

$$y = f(x_1, x_2, \dots, x_n) \quad y \in \mathbb{R}, \quad x_j \in \mathbb{R} \text{ for } j = 1, \dots, n$$

- Example: Black-Scholes Formula for a European Call Option Price

Inputs: S asset price σ asset volatility
 K strike price r risk-free interest rate
 T maturity q asset continuous dividend rate
 t time

$$C(S, t; \cdot) = S e^{-q(T-t)} \Phi(d_+(S, t; \cdot)) - K e^{-r(T-t)} \Phi(d_-(S, t; \cdot))$$

$$d_+(S, t; \cdot) = \frac{\log\left(\frac{S}{K}\right) + (r - q + \frac{\sigma^2}{2})(T - t)}{\sigma\sqrt{T - t}} \quad d_-(S, t; \cdot) = d_+(S, t; \cdot) - \sigma\sqrt{T - t}$$

Partial Derivatives

- Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$
- The partial derivative of f with respect to x_j is denoted by $\frac{\partial f}{\partial x_j}(x_1, \dots, x_n)$ and is defined as

$$\frac{\partial f}{\partial x_j}(\cdot) = \lim_{h \rightarrow 0} \frac{f(x_1, \dots, x_{j-1}, x_j + h, x_{j+1}, \dots, x_n) - f(x_1, \dots, x_n)}{h}$$

if the limit exists and is finite.

- In practice, to compute $\frac{\partial f}{\partial x_j}$
 - fix x_k for $k \neq j$
 - differentiate f as a function of one variable x_j

Example

- $f(x, y) = x^2y + e^{-xy^3}$

$$\frac{\partial}{\partial x} f(x, y) = \frac{\partial}{\partial x} [y x^2] + \frac{\partial}{\partial x} [e^{-xy^3}] \quad \text{let } u = -xy^3$$

$$= y \frac{\partial}{\partial x} [x^2] + \frac{\partial}{\partial x} [e^u]$$

$$= 2xy + e^u \frac{\partial u}{\partial x} \quad \frac{\partial u}{\partial x} = -y^3$$

$$= 2xy - y^3 e^{-xy^3}$$

Example (continued)

- $f(x, y) = x^2y + e^{-xy^3}$

$$\frac{\partial}{\partial y} f(x, y) = \frac{\partial}{\partial y} [y x^2] + \frac{\partial}{\partial y} [e^{-xy^3}] \quad \text{let } u = -xy^3$$

$$= x^2 \frac{\partial}{\partial y} [y] + \frac{\partial}{\partial y} [e^u]$$

$$= x^2 + e^u \frac{\partial u}{\partial y} \quad \frac{\partial u}{\partial y} = -3xy^2$$

$$= x^2 - 3xy^2 e^{-xy^3}$$

The Gradient

- Let $f(x) = f(x_1, \dots, x_n) : \mathbb{R}^n \rightarrow \mathbb{R}$. The gradient of $f(x)$ is denoted by $Df(x)$ and is defined to be the following $1 \times n$ array of partial derivatives.

$$Df(x) = \left[\frac{\partial f}{\partial x_1} \quad \frac{\partial f}{\partial x_2} \quad \cdots \quad \frac{\partial f}{\partial x_n} \right]$$

Vector Valued Functions

- A function of one or several variables that takes values in a multidimensional space is called a vector valued function.
- Notation: $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$

$$f(x_1, \dots, x_n) = \begin{bmatrix} f_1(x_1, \dots, x_n) \\ f_2(x_1, \dots, x_n) \\ \vdots \\ f_m(x_1, \dots, x_n) \end{bmatrix}$$

- Partial derivatives have the form

$$\frac{\partial f_i}{\partial x_j}(x_1, \dots, x_n)$$

- There are $n \times m$ first-order partial derivatives in total. Yikes!

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Higher Order Partial Derivatives

- For functions of a single variable:

$$\frac{d^2}{dx^2}f(x) = \frac{d}{dx} \left[\frac{d}{dx}f(x) \right] = \frac{d}{dx}f'(x) = f''(x)$$

- For functions of several variables:

$$\frac{\partial^2}{\partial x^2}e^{xy} = \frac{\partial}{\partial x} \left[\frac{\partial}{\partial x}e^{xy} \right] = \frac{\partial}{\partial x}[ye^{xy}] = y \frac{\partial}{\partial x}e^{xy} = y^2 e^{xy}$$

$$\frac{\partial^2}{\partial y^2}e^{xy} = \frac{\partial}{\partial y} \left[\frac{\partial}{\partial y}e^{xy} \right] = \frac{\partial}{\partial y}[xe^{xy}] = x \frac{\partial}{\partial y}e^{xy} = x^2 e^{xy}$$

- For functions of several variables also have mixed partial derivatives:

$$\frac{\partial^2}{\partial x \partial y}e^{xy} = \frac{\partial}{\partial x} \left[\frac{\partial}{\partial y}e^{xy} \right] = \frac{\partial}{\partial x}[xe^{xy}] = e^{xy} + x \frac{\partial}{\partial x}e^{xy} = e^{xy} + xy e^{xy}$$

Mixed Partial Derivatives

- What is the relationship between $\frac{\partial^2}{\partial x \partial y}$ and $\frac{\partial^2}{\partial y \partial x}$?

- Already saw that $\frac{\partial^2}{\partial x \partial y} e^{xy} = e^{xy} + xy e^{xy}$

$$\frac{\partial^2}{\partial y \partial x} e^{xy} = \frac{\partial}{\partial y} \left[\frac{\partial}{\partial x} e^{xy} \right] = \frac{\partial}{\partial y} [y e^{xy}] = e^{xy} + y \frac{\partial}{\partial y} e^{xy} = e^{xy} + xy e^{xy}$$

- For $f(x, y) = e^{xy}$ have

$$\frac{\partial^2 f}{\partial x \partial y} = \frac{\partial^2}{\partial x \partial y} e^{xy} = e^{xy} + xy e^{xy} = \frac{\partial^2}{\partial y \partial x} e^{xy} = \frac{\partial^2 f}{\partial y \partial x}$$

- But, $f(x, y) = e^{xy}$ has a certain “*symmetry*” wrt differentiation
- Let’s see what happens when $f(x, y) = x^2 y + e^{-xy^3}$

Mixed Partial Derivatives

$$\frac{\partial^2 f}{\partial x \partial y} = \frac{\partial}{\partial x} \left[\frac{\partial}{\partial y} (x^2 y + e^{-xy^3}) \right] \quad \text{let } u = -xy^3$$

$$= \frac{\partial}{\partial x} \left[x^2 + \frac{\partial}{\partial y} e^u \right]$$

$$= \frac{\partial}{\partial x} \left[x^2 + e^u \frac{\partial u}{\partial y} \right] \quad \frac{\partial u}{\partial y} = -3xy^2$$

$$= \frac{\partial}{\partial x} [x^2 - 3xy^2 e^{-xy^3}]$$

$$= 2x - \left[3y^2 e^{-xy^3} + 3xy^2 \frac{\partial}{\partial x} e^u \right]$$

$$= 2x - 3y^2 e^{-xy^3} - 3xy^2 e^u \frac{\partial u}{\partial x} \quad \frac{\partial u}{\partial x} = -y^3$$

$$= 2x - 3y^2 e^{-xy^3} + 3xy^5 e^{-xy^3}$$

Mixed Partial Derivatives

$$\frac{\partial^2 f}{\partial y \partial x} = \frac{\partial}{\partial y} \left[\frac{\partial}{\partial x} (x^2 y + e^{-xy^3}) \right] \quad \text{let } u = -xy^3$$

$$= \frac{\partial}{\partial y} \left[2xy + \frac{\partial}{\partial x} e^u \right]$$

$$= \frac{\partial}{\partial y} \left[2xy + e^u \frac{\partial u}{\partial x} \right] \quad \frac{\partial u}{\partial x} = -y^3$$

$$= \frac{\partial}{\partial y} [2xy - y^3 e^u]$$

$$= 2x - \left[3y^2 e^{-xy^3} + y^3 \frac{\partial}{\partial y} e^u \right]$$

$$= 2x - 3y^2 e^{-xy^3} - y^3 e^u \frac{\partial u}{\partial y} \quad \frac{\partial u}{\partial y} = -3xy^2$$

$$= 2x - 3y^2 e^{-xy^3} + 3xy^5 e^{-xy^3}$$

Mixed Partial Derivatives

- When $f(x, y) = x^2y + e^{-xy^3}$ have

$$\frac{\partial^2 f}{\partial x \partial y} = 2x - 3y^2 e^{-xy^3} + 3xy^5 e^{-xy^3} = \frac{\partial^2 f}{\partial y \partial x}$$

- **Theorem** If all of the partial derivatives of order k of the function $f(x)$ exist and are continuous, then the order in which partial derivatives of $f(x)$ of order at most k are computed does not matter.

The Hessian

- Let $f(x) = f(x_1, \dots, x_n) : \mathbb{R}^n \rightarrow \mathbb{R}$. The Hessian of $f(x)$ is denoted by $D^2f(x)$ and is defined to be the following $n \times n$ array of (mixed) partial derivatives.

$$D^2f(x) = \begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2} & \frac{\partial^2 f}{\partial x_2 \partial x_1} & \cdots & \frac{\partial^2 f}{\partial x_n \partial x_1} \\ \frac{\partial^2 f}{\partial x_1 \partial x_2} & \frac{\partial^2 f}{\partial x_2^2} & \cdots & \frac{\partial^2 f}{\partial x_n \partial x_2} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial^2 f}{\partial x_1 \partial x_n} & \frac{\partial^2 f}{\partial x_2 \partial x_n} & \cdots & \frac{\partial^2 f}{\partial x_n^2} \end{bmatrix}$$

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Functions of Two Variables

- Let $f = f(x, y) : \mathbb{R}^2 \rightarrow \mathbb{R}$ be a scalar-valued function
- The partial derivatives of f are also functions of x and y :

$$\frac{\partial f}{\partial x}(x, y) = \lim_{h \rightarrow 0} \frac{f(x + h, y) - f(x, y)}{h}$$

$$\frac{\partial f}{\partial y}(x, y) = \lim_{h \rightarrow 0} \frac{f(x, y + h) - f(x, y)}{h}$$

- The gradient of $f(x, y)$ is

$$Df(x, y) = \left[\frac{\partial f}{\partial x}(x, y) \quad \frac{\partial f}{\partial y}(x, y) \right]$$

- The Hessian of $f(x, y)$ is

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

Example

- Let $f(x, y) = x^2y^3$. Evaluate Df and D^2f at the point $(1, 2)$

$$\frac{\partial f}{\partial x}(x, y) = 2xy^3$$

$$\frac{\partial f}{\partial y}(x, y) = 3x^2y^2$$

$$\frac{\partial^2 f}{\partial x^2}(x, y) = \frac{\partial}{\partial x} \left[\frac{\partial f}{\partial x}(x, y) \right] = \frac{\partial}{\partial x} [2xy^3] = 2y^3$$

$$\frac{\partial^2 f}{\partial x \partial y}(x, y) = \frac{\partial}{\partial x} \left[\frac{\partial f}{\partial y}(x, y) \right] = \frac{\partial}{\partial x} [3x^2y^2] = 6xy^2 = \frac{\partial^2 f}{\partial y \partial x}(x, y)$$

$$\frac{\partial^2 f}{\partial y^2}(x, y) = \frac{\partial}{\partial y} \left[\frac{\partial f}{\partial y}(x, y) \right] = \frac{\partial}{\partial y} [3x^2y^2] = 6x^2y$$

Example (continued)

$$Df(x, y) = \left[\frac{\partial f}{\partial x}(x, y) \quad \frac{\partial f}{\partial y}(x, y) \right] = [2xy^3 \quad 3x^2y^2]$$

$$Df(1, 2) = [2 \times 1 \times 2^3 \quad 3 \times 1^2 \times 2^2] = [16 \quad 12]$$

$$D^2f(x, y) = \begin{bmatrix} \frac{\partial^2 f}{\partial x^2} & \frac{\partial^2 f}{\partial y \partial x} \\ \frac{\partial^2 f}{\partial x \partial y} & \frac{\partial^2 f}{\partial y^2} \end{bmatrix} = \begin{bmatrix} 2y^3 & 6xy^2 \\ 6xy^2 & 6x^2y \end{bmatrix}$$

$$D^2f(1, 2) = \begin{bmatrix} 2 \times 2^3 & 6 \times 1 \times 2^2 \\ 6 \times 1 \times 2^2 & 6 \times 1^2 \times 2 \end{bmatrix} = \begin{bmatrix} 16 & 24 \\ 24 & 12 \end{bmatrix}$$

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Chain Rule for Functions of a Single Variable

- Let $f(x)$ be a differentiable function
- Let $x = g(t)$ where $g(t)$ is a differentiable function
- $f(x)$ can be thought of as a function of t : $f(x) = f(g(t))$, and

$$\frac{df}{dt} = \frac{df}{dx} \frac{dx}{dt}$$

- Example: evaluate $\frac{d}{dt} \log(\cos(t))$

Let $f(x) = \log(x)$, $x = \cos(t)$

$$\frac{df}{dt} = \frac{d}{dt} f(x) = \frac{df}{dx} \frac{dx}{dt} = \frac{1}{x} [-\sin(t)] = -\frac{1}{\cos(t)} \sin(t) = -\tan(t)$$

Chain Rule for Functions of 2 Variables

- Let $f(x, y)$ be a differentiable function
- Let $x = g(t)$ and $y = h(t)$ where g and h are differentiable functions
- $f(x, y) = f(g(t), h(t))$ is a function of t and

$$\frac{df}{dt}(g(t), h(t)) = \frac{\partial f}{\partial x}(g(t), h(t)) g'(t) + \frac{\partial f}{\partial y}(g(t), h(t)) h'(t)$$

- Using Leibniz notation:

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

Example

- Let $f(x, y) = x^2 + y + xy^3$, $x = e^{2t}$, and $y = t^2$
- First, by direct computation

$$f(t) = e^{4t} + t^2 + t^6 e^{2t}$$

$$\frac{d}{dt}f(t) = 4e^{4t} + 2t + [6t^5 e^{2t} + t^6 2e^{2t}]$$

$$\frac{df}{dt} = 2t^6 e^{2t} + 6t^5 e^{2t} + 2t + 4e^{4t}$$

Example (continued)

- Again, using the chain rule

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

$$\frac{\partial f}{\partial x} = 2x + y^3 \qquad \frac{\partial f}{\partial y} = 1 + 3xy^2$$

$$\frac{dx}{dt} = 2e^{2t} \qquad \frac{dy}{dt} = 2t$$

$$\begin{aligned} \frac{d}{dt}f(x, y) &= \frac{\partial f}{\partial x} \frac{dx}{dt} + \frac{\partial f}{\partial y} \frac{dy}{dt} = (2x + y^3) 2e^{2t} + (1 + 3xy^2) 2t \\ &= (2e^{2t} + t^6) 2e^{2t} + (1 + 3t^4 e^{2t}) 2t \\ &= 2t^6 e^{2t} + 6t^5 e^{2t} + 2t + 4e^{4t} \end{aligned}$$

Chain Rule for Functions of 2 Variables

- Let $f(x, y)$ be a differentiable function
- Let $x = g(s, t)$ and $y = h(s, t)$ where g and h are differentiable
- $f(x, y) = f(g(s, t), h(s, t))$ is a function of s and t

$$\begin{aligned}\frac{\partial f}{\partial s} &= \frac{\partial f}{\partial x} \frac{\partial x}{\partial s} + \frac{\partial f}{\partial y} \frac{\partial y}{\partial s} \\ &= \frac{\partial f}{\partial x} \frac{\partial g}{\partial s} + \frac{\partial f}{\partial y} \frac{\partial h}{\partial s}\end{aligned}$$

$$\begin{aligned}\frac{\partial f}{\partial t} &= \frac{\partial f}{\partial x} \frac{\partial x}{\partial t} + \frac{\partial f}{\partial y} \frac{\partial y}{\partial t} \\ &= \frac{\partial f}{\partial x} \frac{\partial g}{\partial t} + \frac{\partial f}{\partial y} \frac{\partial h}{\partial t}\end{aligned}$$

Chain Rule for Functions of n Variables

- In general, let $f = f(x_1, \dots, x_n)$ be a function of n variables
- For $i = 1, \dots, n$, let $x_i = x_i(t_1, \dots, t_m)$ be functions of m variables
- The partial derivative of f wrt t_j is

$$\frac{\partial f}{\partial t_j} = \sum_{i=1}^n \frac{\partial f}{\partial x_i} \frac{\partial x_i}{\partial t_j}$$

Example

- $f(x_1, x_2, x_3) = x_1^2 + x_1x_2 + x_1x_3 + 2x_3^2$
 $x_1(t_1, t_2) = t_1^2 - t_2^2 + 1, \quad x_2(t_1, t_2) = t_2^2 + t_1 + 1, \quad x_3(t_1, t_2) = -t_1^2 - 1$

- Compute $\frac{\partial f}{\partial t_1}$

$$\begin{aligned}\frac{\partial f}{\partial t_1} &= \frac{\partial f}{\partial x_1} \frac{\partial x_1}{\partial t_1} + \frac{\partial f}{\partial x_2} \frac{\partial x_2}{\partial t_1} + \frac{\partial f}{\partial x_3} \frac{\partial x_3}{\partial t_1} \\&= (2x_1 + x_2 + x_3)(2t_1) + (x_1)(1) + (x_1 + 4x_3)(-2t_1) \\&= (2t_1^2 - 2t_2^2 + 2 + t_2^2 + t_1 + 1 - t_1^2 - 1)(2t_1) \\&\quad + (t_1^2 - t_2^2 + 1) + (t_1^2 - t_2^2 + 1 - 4t_1^2 - 4)(-2t_1) \\&= (t_1^2 - t_2^2 + t_1 + 2)(2t_1) + (t_1^2 - t_2^2 + 1) + (3t_1^2 + t_2^2 + 3)(2t_1) \\&= 8t_1^3 + 3t_1^2 + 10t_1 - t_2^2 + 1\end{aligned}$$

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Implicit Functions

- So far, functions have expressed one variable in terms of another (or others), e.g.,

$$y = f(x) = \sqrt{1 - x^2} \quad \text{or} \quad y = \frac{\sin(x)}{x}$$

- An implicit function is defined by a more general relation between the variables, e.g.,

$$x^2 + y^2 = 1 \quad \text{or} \quad x^3 + y^3 = 6xy$$

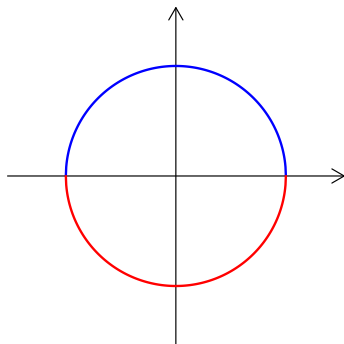
- In some cases, possible to solve for one variable as an explicit function (or functions) of the others.
- Terminology: let $F(x, y, z)$ be a function of 3 variables. The set of points that satisfy

$$F(x, y, z) = 0$$

is called the locus defined by F .

Implicit Functions

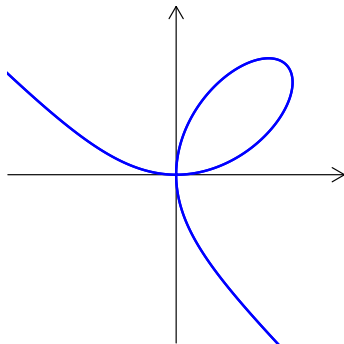
Circle



$$x^2 + y^2 = 1$$

- blue curve: $y = \sqrt{1 - x^2}$
- red curve: $y = -\sqrt{1 - x^2}$

Folium



$$x^3 + y^3 = 6xy$$

- blue curve: more difficult ...

Circle

Slope of tangent line to the unit circle

- Case 1: blue curve $y \geq 0$

$$\frac{d}{dx}y = \frac{d}{dx}\sqrt{1-x^2}$$

$$\frac{dy}{dx} = \frac{d}{dx}(1-x^2)^{\frac{1}{2}}$$

$$= \frac{1}{2}(1-x^2)^{-\frac{1}{2}}(-2x)$$

$$= \frac{-x}{\sqrt{1-x^2}}$$

- Case 2: red curve $y < 0$

$$\frac{d}{dx}y = \frac{d}{dx}\left[-\sqrt{1-x^2}\right]$$

$$\frac{dy}{dx} = \frac{d}{dx}\left[-(1-x^2)^{\frac{1}{2}}\right]$$

$$= -\frac{1}{2}(1-x^2)^{-\frac{1}{2}}(-2x)$$

$$= \frac{-x}{-\sqrt{1-x^2}}$$

Derivatives of Implicit Functions Using the Chain Rule

- Recall that if y is a function of x then the chain rule says

$$\frac{d}{dx}y^2 = \frac{d}{dy}[y^2] \frac{dy}{dx} = 2y \frac{dy}{dx}$$

$$x^2 + y^2 = 1$$

$$2y \frac{dy}{dx} = -2x$$

$$\frac{d}{dx}[x^2 + y^2] = \frac{d}{dx} 1$$

$$\frac{dy}{dx} = \frac{-x}{y}$$

$$\frac{d}{dx}x^2 + \frac{d}{dx}y^2 = 0$$

$$= \frac{-x}{\sqrt{1-x^2}} \quad (y \geq 0)$$

$$2x + 2y \frac{dy}{dx} = 0$$

$$= \frac{-x}{-\sqrt{1-x^2}} \quad (y < 0)$$

Example

- Compute $\frac{dy}{dx}$ for the Folium

$$x^3 + y^3 = 6xy$$

$$\frac{d}{dx}[x^3 + y^3] = \frac{d}{dx}[6xy]$$

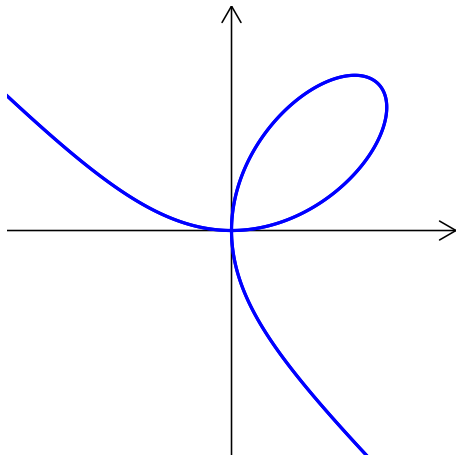
$$\frac{d}{dx}[x^3] + \frac{d}{dx}[y^3] = 6y + 6x \frac{d}{dx}[y]$$

$$3x^2 + 3y^2 \frac{dy}{dx} = 6y + 6x \frac{dy}{dx}$$

$$(3y^2 - 6x) \frac{dy}{dx} = 6y - 3x^2$$

$$\frac{dy}{dx} = \frac{2y - x^2}{y^2 - 2x}$$

Example (continued)



$$\frac{dy}{dx} = \frac{2y - x^2}{y^2 - 2x}$$

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The Greeks

- Let V be the value of a portfolio of derivative securities based on one underlying asset
- The rates of change of the value V wrt pricing parameters (e.g., asset price, volatility, etc.) useful for hedging
- These rates of change are called the Greeks of the portfolio
- Consider a portfolio containing a single European call option
- Price using Black-Scholes

Inputs:	S	asset price	σ	asset volatility
	K	strike price	r	(continuous) risk-free interest rate
	T	maturity	q	(continuous) asset dividend rate
	t	time		

- Black-Scholes formula for a European call option:

$$C(S, t) = Se^{-q(T-t)}\Phi(d_+) - Ke^{-r(T-t)}\Phi(d_-)$$

where

$$\Phi(z) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^z e^{-\frac{x^2}{2}} dx$$

$$d_+ = \frac{\log\left(\frac{S}{K}\right) + \left(r - q + \frac{\sigma^2}{2}\right)(T-t)}{\sigma\sqrt{T-t}}$$

$$d_- = d_+ - \sigma\sqrt{T-t} = \frac{\log\left(\frac{S}{K}\right) + \left(r - q - \frac{\sigma^2}{2}\right)(T-t)}{\sigma\sqrt{T-t}}$$

Put-Call Parity

- $C(t)$ and $P(t)$ prices of European call and put options on same asset
- Same maturity T and strike price K
- Put-Call parity states that

$$P(t) + S(t)e^{-q(T-t)} - C(t) = Ke^{-r(T-t)}$$

The Greeks

- Delta (Δ): rate of change of C wrt S $\Delta = \frac{\partial C}{\partial S}$
- Gamma (Γ): rate of change of Δ wrt S $\Gamma = \frac{\partial \Delta}{\partial S} = \frac{\partial^2 C}{\partial S^2}$
- Theta (Θ): rate of change of C wrt t $\Theta = \frac{\partial C}{\partial t}$
- Rho (ρ): rate of change of C wrt r $\rho = \frac{\partial C}{\partial r}$
- Vega: rate of change of C wrt σ $\text{vega} = \frac{\partial C}{\partial \sigma}$

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- The Delta (Δ) of a European call option is the rate of change of $C(S, t)$ wrt the asset price S

$$\begin{aligned}\Delta &= \frac{\partial}{\partial S} C(S, t) \\&= \frac{\partial}{\partial S} \left[S e^{-q(T-t)} \Phi(d_+) - K e^{-r(T-t)} \Phi(d_-) \right] \\&= e^{-q(T-t)} \frac{\partial}{\partial S} [S \Phi(d_+)] - K e^{-r(T-t)} \frac{\partial}{\partial S} [\Phi(d_-)]\end{aligned}$$

- By the product rule:

$$= e^{-q(T-t)} \left[\Phi(d_+) + S \frac{\partial}{\partial S} \Phi(d_+) \right] - K e^{-r(T-t)} \frac{\partial}{\partial S} [\Phi(d_-)]$$

Delta (continued)

- Consider the partial derivatives

$$\frac{\partial}{\partial S} [\Phi(d_{\pm})]$$

- The chain rule says

$$\frac{\partial}{\partial S} [\Phi(d_{\pm})] = \frac{\partial}{\partial d_{\pm}} [\Phi(d_{\pm})] \frac{\partial d_{\pm}}{\partial S}$$

- Recall definition of Φ

$$\Phi(d_{\pm}) = \int_{-\infty}^{d_{\pm}} \phi(x) dx = \int_{-\infty}^{d_{\pm}} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx$$

$$\frac{\partial}{\partial d_{\pm}} \Phi(d_{\pm}) = \frac{\partial}{\partial d_{\pm}} \int_{-\infty}^{d_{\pm}} \phi(x) dx = \frac{\partial}{\partial d_{\pm}} \int_{-\infty}^{d_{\pm}} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx$$

$$\frac{\partial}{\partial d_{\pm}} \Phi(d_{\pm}) = \phi(d_{\pm}) = \frac{1}{\sqrt{2\pi}} e^{-\frac{(d_{\pm})^2}{2}}$$

Delta (continued)

- Results so far

$$\Delta = e^{-q(T-t)} \left[\Phi(d_+) + S \frac{\partial}{\partial S} \Phi(d_+) \right] - Ke^{-r(T-t)} \frac{\partial}{\partial S} [\Phi(d_-)]$$

$$\frac{\partial}{\partial S} [\Phi(d_{\pm})] = \frac{\partial}{\partial d_{\pm}} [\Phi(d_{\pm})] \frac{\partial d_{\pm}}{\partial S}$$

$$\frac{\partial}{\partial d_{\pm}} \Phi(d_{\pm}) = \phi(d_{\pm}) = \frac{1}{\sqrt{2\pi}} e^{-\frac{(d_{\pm})^2}{2}}$$

- Substituting

$$\Delta = e^{-q(T-t)} \left[\Phi(d_+) + S \phi(d_+) \frac{\partial d_+}{\partial S} \right] - Ke^{-r(T-t)} \phi(d_-) \frac{\partial d_-}{\partial S}$$

Delta (continued)

- Finally, need to compute partial derivatives of d_+ and d_- wrt to S

$$d_{\pm} = \frac{\log\left(\frac{S}{K}\right) + \left(r - q \pm \frac{\sigma^2}{2}\right)(T - t)}{\sigma\sqrt{T - t}}$$

$$\frac{\partial}{\partial S} d_{\pm} = \frac{1}{\sigma\sqrt{T - t}} \frac{\partial}{\partial S} \left[\log\left(\frac{S}{K}\right) \right] + \frac{\partial}{\partial S} \left[\frac{\left(r - q \pm \frac{\sigma^2}{2}\right)(T - t)}{\sigma\sqrt{T - t}} \right]$$

- Let $u = \frac{S}{K}$, then by the chain rule

$$\begin{aligned} \frac{\partial}{\partial S} d_{\pm} &= \frac{1}{\sigma\sqrt{T - t}} \frac{\partial}{\partial S} [\log(u)] = \frac{1}{\sigma\sqrt{T - t}} \frac{1}{u} \frac{\partial u}{\partial S} = \frac{1}{\sigma\sqrt{T - t}} \frac{K}{S} \frac{1}{K} \\ &= \frac{1}{S\sigma\sqrt{T - t}} \end{aligned}$$

Delta (continued)

- Putting it all together ...

$$\Delta = e^{-q(T-t)} \left[\Phi(d_+) + S \phi(d_+) \frac{\partial d_+}{\partial S} \right] - K e^{-r(T-t)} \phi(d_-) \frac{\partial d_-}{\partial S}$$

$$\frac{\partial}{\partial S} d_{\pm} = \frac{1}{S \sigma \sqrt{T-t}}$$

- Yields the following expression for Δ

$$\Delta = e^{-q(T-t)} \Phi(d_+) + \frac{e^{-q(T-t)} \phi(d_+)}{\sigma \sqrt{T-t}} - \frac{K e^{-r(T-t)} \phi(d_-)}{S \sigma \sqrt{T-t}}$$

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Gamma (Γ)

- Gamma (Γ) is the rate of change of Delta (Δ)
- Hence: Gamma (Γ) is the second partial derivative of C wrt S

$$\Gamma = \frac{\partial}{\partial S} \Delta = \frac{\partial}{\partial S} \frac{\partial C}{\partial S} = \frac{\partial^2 C}{\partial S^2}$$

- So, all we have to do is ...

$$\Gamma = \frac{\partial}{\partial S} \Delta = \frac{\partial}{\partial S} \left[e^{-q(T-t)} \Phi(d_+) + \frac{e^{-q(T-t)} \phi(d_+)}{\sigma \sqrt{T-t}} - \frac{Ke^{-r(T-t)} \phi(d_-)}{S \sigma \sqrt{T-t}} \right]$$

- Luckily, there is a shortcut

Simplifying the Expression for Delta

- The textbook says

$$\Delta = e^{-q(T-t)}\phi(d_+) + \frac{e^{-q(T-t)}\phi(d_+)}{\sigma\sqrt{T-t}} - \frac{Ke^{-r(T-t)}\phi(d_-)}{S\sigma\sqrt{T-t}}$$

- Finding an expression for Γ much easier if

$$\frac{e^{-q(T-t)}\phi(d_+)}{\sigma\sqrt{T-t}} - \frac{Ke^{-r(T-t)}\phi(d_-)}{S\sigma\sqrt{T-t}} \stackrel{?}{=} 0$$

- Strategy: manipulate $\phi(d_-)$ into $\phi(d_+)$ and see what falls out

$$\phi(d_-) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{d_-^2}{2}\right) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{(d_+ - \sigma\sqrt{T-t})^2}{2}\right)$$

Simplifying the Expression for Delta

$$\begin{aligned}\phi(d_-) &= \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{(d_+ - \sigma\sqrt{T-t})^2}{2}\right) \\&= \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{d_+^2 - 2d_+ \sigma\sqrt{T-t} + \sigma^2(T-t)}{2}\right) \\&= \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{d_+^2}{2}\right) \exp\left(-\frac{-2d_+ \sigma\sqrt{T-t}}{2}\right) \exp\left(-\frac{\sigma^2(T-t)}{2}\right) \\&= \phi(d_+) \exp\left(d_+ \sigma\sqrt{T-t}\right) \exp\left(-\sigma^2(T-t)/2\right)\end{aligned}$$

Reminder:
$$d_+ = \frac{\log\left(\frac{S}{K}\right) + \left(r - q + \frac{\sigma^2}{2}\right)(T-t)}{\sigma\sqrt{T-t}}$$

$$= \phi(d_+) \exp\left[\log\left(\frac{S}{K}\right) + \left(r - q + \frac{\sigma^2}{2}\right)(T-t)\right] \exp\left[-\frac{\sigma^2(T-t)}{2}\right]$$

Simplifying the Expression for Delta

$$\begin{aligned}\phi(d_-) &= \phi(d_+) \frac{S}{K} \exp \left[\left(r - q + \frac{\sigma^2}{2} \right) (T - t) - \frac{\sigma^2}{2} (T - t) \right] \\ &= \phi(d_+) \frac{S}{K} e^{r(T-t)} e^{-q(T-t)}\end{aligned}$$

- Substituting

$$\begin{aligned}& \frac{e^{-q(T-t)} \phi(d_+)}{\sigma \sqrt{T-t}} - \frac{K e^{-r(T-t)} \phi(d_-)}{S \sigma \sqrt{T-t}} \\ &= \frac{e^{-q(T-t)} \phi(d_+)}{\sigma \sqrt{T-t}} - \frac{K e^{-r(T-t)} \phi(d_+) \frac{S}{K} e^{r(T-t)} e^{-q(T-t)}}{S \sigma \sqrt{T-t}} \\ &= \frac{e^{-q(T-t)} \phi(d_+)}{\sigma \sqrt{T-t}} - \frac{e^{-q(T-t)} \phi(d_+)}{\sigma \sqrt{T-t}} = 0\end{aligned}$$

Gamma (Γ)

$$\begin{aligned}\Gamma = \frac{\partial}{\partial S} \Delta &= \frac{\partial}{\partial S} \left[e^{-q(T-t)} \Phi(d_+) + \frac{e^{-q(T-t)} \phi(d_+)}{\sigma \sqrt{T-t}} - \frac{K e^{-r(T-t)} \phi(d_-)}{S \sigma \sqrt{T-t}} \right] \\&= \frac{\partial}{\partial S} \left[e^{-q(T-t)} \Phi(d_+) \right] \\&= e^{-q(T-t)} \frac{\partial}{\partial S} [\Phi(d_+)] \\&= e^{-q(T-t)} \frac{\partial}{\partial d_+} \Phi(d_+) \frac{\partial d_+}{\partial S} \qquad \frac{\partial d_{\pm}}{\partial S} = \frac{1}{S \sigma \sqrt{T-t}} \\&= \frac{e^{-q(T-t)}}{S \sigma \sqrt{T-t}} \phi(d_+)\end{aligned}$$

$$\Gamma = \frac{\partial}{\partial S} \Delta = \frac{e^{-q(T-t)}}{S \sigma \sqrt{T-t}} \frac{1}{\sqrt{2\pi}} e^{-\frac{d_+^2}{2}}$$

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Rho (ρ)

- Rho (ρ) is the rate of change of the value of the portfolio wrt the risk-free interest rate r
- Black-Scholes formula

$$C(S, t) = Se^{-q(T-t)}\Phi(d_+) - Ke^{-r(T-t)}\Phi(d_-)$$

- Derivative wrt r :

$$\rho = \frac{\partial}{\partial r} \left[Se^{-q(T-t)}\Phi(d_+) - Ke^{-r(T-t)}\Phi(d_-) \right]$$

- Product rule:

$$\begin{aligned} \rho = Se^{-q(T-t)} \frac{\partial}{\partial r} \Phi(d_+) \\ - \left[-K(T-t)e^{-r(T-t)}\Phi(d_-) + Ke^{-r(T-t)} \frac{\partial}{\partial r} \Phi(d_-) \right] \end{aligned}$$

Rho (ρ)

- Chain rule:

$$\rho = Se^{-q(T-t)}\phi(d_+)\frac{\partial d_+}{\partial r} - \left[-K(T-t)e^{-r(T-t)}\Phi(d_-) + Ke^{-r(T-t)}\phi(d_-)\frac{\partial d_-}{\partial r} \right]$$

- Need partial derivatives of d_+ and d_- wrt r :

$$d_{\pm} = \frac{\log\left(\frac{S}{K}\right) + \left(r - q \pm \frac{\sigma^2}{2}\right)(T-t)}{\sigma\sqrt{T-t}} = \frac{r\sqrt{T-t}}{\sigma} + \mathcal{C}$$

$$\frac{\partial}{\partial r}d_{\pm} = \frac{\sqrt{T-t}}{\sigma}$$

Rho (ρ)

- Substituting $\frac{\partial}{\partial r} d_{\pm} = \frac{\sqrt{T-t}}{\sigma} \dots$

$$\begin{aligned} \rho = & K(T-t)e^{-r(T-t)} \Phi(d_-) \\ & + Se^{-q(T-t)} \phi(d_+) \frac{\sqrt{T-t}}{\sigma} + Ke^{-r(T-t)} \phi(d_-) \frac{\sqrt{T-t}}{\sigma} \end{aligned}$$

- Rewrite as ...

$$\begin{aligned} \rho = & K(T-t)e^{-r(T-t)} \Phi(d_-) \\ & + S(T-t) \left[\frac{e^{-q(T-t)} \phi(d_+)}{\sigma \sqrt{T-t}} + \frac{Ke^{-r(T-t)} \phi(d_-)}{S \sigma \sqrt{T-t}} \right] \end{aligned}$$

$$\rho = K(T-t)e^{-r(T-t)} \Phi(d_-)$$

- Vega is the rate of change of the value of the portfolio wrt the volatility σ
- Black-Scholes formula

$$C(S, t) = Se^{-q(T-t)}\Phi(d_+) - Ke^{-r(T-t)}\Phi(d_-)$$

- Derivative wrt σ :

$$\text{vega} = \frac{\partial}{\partial \sigma} \left[Se^{-q(T-t)}\Phi(d_+) - Ke^{-r(T-t)}\Phi(d_-) \right]$$

- Chain rule:

$$\text{vega} = Se^{-q(T-t)}\phi(d_+) \frac{\partial d_+}{\partial \sigma} - Ke^{-r(T-t)}\phi(d_-) \frac{\partial d_-}{\partial \sigma}$$

- Partial derivatives of d_+ and d_- wrt σ :

$$d_{\pm} = \frac{\log\left(\frac{S}{K}\right) + \left(r - q \pm \frac{\sigma^2}{2}\right)(T - t)}{\sigma\sqrt{T - t}}$$

$$= \frac{\log\left(\frac{S}{K}\right) + (r - q)(T - t)}{\sqrt{T - t}}\sigma^{-1} \pm \frac{(T - t)}{2\sqrt{T - t}}\sigma$$

$$\frac{\partial d_{\pm}}{\partial \sigma} = -\frac{\log\left(\frac{S}{K}\right) + (r - q)(T - t)}{\sqrt{T - t}}\sigma^{-2} \pm \frac{(T - t)}{2\sqrt{T - t}}$$

$$= -\frac{\log\left(\frac{S}{K}\right) + (r - q)(T - t)}{\sigma^2\sqrt{T - t}} \pm \frac{\frac{\sigma^2}{2}(T - t)}{\sigma^2\sqrt{T - t}}$$

$$= -\frac{\log\left(\frac{S}{K}\right) + \left(r - q \mp \frac{\sigma^2}{2}\right)(T - t)}{\sigma^2\sqrt{T - t}}$$

Bag of Tricks

- Substitute

$$\frac{\partial}{\partial \sigma} d_{\pm} = - \frac{\log\left(\frac{S}{K}\right) + (r - q \mp \frac{\sigma^2}{2})(T - t)}{\sigma^2 \sqrt{T - t}}$$

into

$$\text{vega} = S e^{-q(T-t)} \phi(d_+) \frac{\partial d_+}{\partial \sigma} - K e^{-r(T-t)} \phi(d_-) \frac{\partial d_-}{\partial \sigma}$$

- Looks like it's going to get worse before it gets better
- Another approach ...
- Already have

$$\frac{e^{-q(T-t)} \phi(d_+)}{\sigma \sqrt{T-t}} - \frac{K e^{-r(T-t)} \phi(d_-)}{S \sigma \sqrt{T-t}} = 0$$

- Rewrite as

$$S e^{-q(T-t)} \phi(d_+) = K e^{-r(T-t)} \phi(d_-)$$

- The expression for vega becomes

$$\begin{aligned}\text{vega} &= Se^{-q(T-t)}\phi(d_+)\frac{\partial d_+}{\partial\sigma} - Ke^{-r(T-t)}\phi(d_-)\frac{\partial d_-}{\partial\sigma} \\&= Se^{-q(T-t)}\phi(d_+)\left(\frac{\partial d_+}{\partial\sigma} - \frac{\partial d_-}{\partial\sigma}\right) \\&= Se^{-q(T-t)}\phi(d_+)\frac{\partial}{\partial\sigma}(d_+ - d_-)\end{aligned}$$

- Recall: $d_- = d_+ - \sigma\sqrt{T-t}$ so that

$$\begin{aligned}d_+ - d_- &= d_+ - (d_+ - \sigma\sqrt{T-t}) \\ \frac{\partial}{\partial\sigma}(d_+ - d_-) &= \frac{\partial}{\partial\sigma}\sigma\sqrt{T-t} \\ &= \sqrt{T-t}\end{aligned}$$

- Substituting ...

$$\text{vega} = S e^{-q(T-t)} \phi(d_+) \frac{\partial}{\partial \sigma} (d_+ - d_-)$$

$$\text{vega} = S \sqrt{T-t} e^{-q(T-t)} \phi(d_+)$$

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Theta (Θ)

- Theta is the rate of change of the value of the portfolio wrt the time t
- Black-Scholes formula

$$C(S, t) = Se^{-q(T-t)}\Phi(d_+) - Ke^{-r(T-t)}\Phi(d_-)$$

- Derivative wrt t :

$$\Theta = \frac{\partial}{\partial t} \left[Se^{-q(T-t)}\Phi(d_+) - Ke^{-r(T-t)}\Phi(d_-) \right]$$

- Product rule twice:

$$\begin{aligned} \Theta &= qSe^{-q(T-t)}\Phi(d_+) + Se^{-q(T-t)}\frac{\partial}{\partial t}\Phi(d_+) \\ &\quad - \left[rKe^{-r(T-t)}\Phi(d_-) + Ke^{-r(T-t)}\frac{\partial}{\partial t}\Phi(d_-) \right] \end{aligned}$$

Theta (Θ)

$$\begin{aligned}\Theta &= qSe^{-q(T-t)}\Phi(d_+) - rKe^{-r(T-t)}\Phi(d_-) \\ &\quad + \left[Se^{-q(T-t)}\frac{\partial}{\partial t}\Phi(d_+) - Ke^{-r(T-t)}\frac{\partial}{\partial t}\Phi(d_-) \right] \\ &= qSe^{-q(T-t)}\Phi(d_+) - rKe^{-r(T-t)}\Phi(d_-) \\ &\quad + \left[Se^{-q(T-t)}\phi(d_+)\frac{\partial d_+}{\partial t} - Ke^{-r(T-t)}\phi(d_-)\frac{\partial d_-}{\partial t} \right] \\ &= qSe^{-q(T-t)}\Phi(d_+) - rKe^{-r(T-t)}\Phi(d_-) \\ &\quad + Se^{-q(T-t)}\phi(d_+)\left(\frac{\partial d_+}{\partial t} - \frac{\partial d_-}{\partial t}\right)\end{aligned}$$

Theta (Θ)

- Again, rewrite the quantity

$$\begin{aligned}\left(\frac{\partial d_+}{\partial t} - \frac{\partial d_-}{\partial t}\right) &= \frac{\partial}{\partial t}(d_+ - d_-) \\ &= \frac{\partial}{\partial t}\sigma\sqrt{T-t} = \sigma\frac{\partial}{\partial t}(T-t)^{\frac{1}{2}} \\ &= \frac{-\sigma}{2\sqrt{T-t}}\end{aligned}$$

- Substitute into ...

$$\begin{aligned}\Theta &= qSe^{-q(T-t)}\Phi(d_+) - rKe^{-r(T-t)}\Phi(d_-) \\ &\quad + Se^{-q(T-t)}\phi(d_+)\left(\frac{\partial d_+}{\partial t} - \frac{\partial d_-}{\partial t}\right)\end{aligned}$$

Theta (Θ)

- Finally ...

$$\Theta = -\frac{\sigma S e^{-q(T-t)}}{2\sqrt{T-t}} \phi(d_+) + q S e^{-q(T-t)} \Phi(d_+) - r K e^{-r(T-t)} \Phi(d_-)$$



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