

On the modelling of steady, inviscid and incompressible fluid flow around a two-dimensional cylinder

Research question: "How can vector calculus be used to model and analyse steady, inviscid, and incompressible fluid flow in two-dimensional spaces around a circular obstacle, and what mathematical principles underpin the observed fluid behaviour?"

Mathematics AA HL

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1 Introduction

Fluid dynamics is today a cornerstone to several fields of study, including aerospace engineering and meteorology. Real world fluid behaviour is intricate and complex. Therefore, to gain insights into the governing principles of fluid flow, simplified and idealised models are used. This essay investigates the application of vector calculus to model and analyse steady, inviscid, and incompressible fluid flow in two-dimensional spaces around a circular obstacle. These idealisations allow for the derivation of some of fluid dynamic's key mathematical formulæ and provides a foundation for understanding less idealised fluids.

This essay will address the question: "How can vector calculus be used to model and analyse steady, inviscid, and incompressible fluid flow in two-dimensional spaces around a circular obstacle, and what mathematical principles underpin the observed fluid behaviour?" Through the derivation of the velocity potential and vector field, this essay aims to demonstrate how fundamental laws of fluid motion can be expressed and used through vector calculus.

1.1 Aim & scope

The scope of this essay will be limited to the theoretical modelling of fluid flow in a two-dimensional space as a vector field under idealised conditions forming steady, inviscid and incompressible fluid flow through the derivation of the velocity-potential. The analysis will be centred on the application of vector calculus to derive key equations and describe the behaviour of idealised fluid around a stationary circular obstacle. Consequently, this essay will not touch on viscous effects, turbulent flow or three-dimensional analysis, nor will it involve any experimental validation. The focus is on the mathematical derivation and analysis of the idealised model.

1.2 Background

1.2.1 Glossary

Definition 1.1. *Steady flow* refers to flow in which the velocity at every point does not change over time [CRACIUNOIU and CIOCIRLAN, 2001].

Definition 1.2. *Inviscid flow* is the flow of a fluid with 0 viscosity [Anderson, 2003].

Definition 1.3. An *incompressible fluid* is a fluid whose density at every point does not change over time [Ahmed, 2019].

Definition 1.4. A *scalar field* is a function mapping points in space to scalar quantities such as temperatures^[see Figure 1].

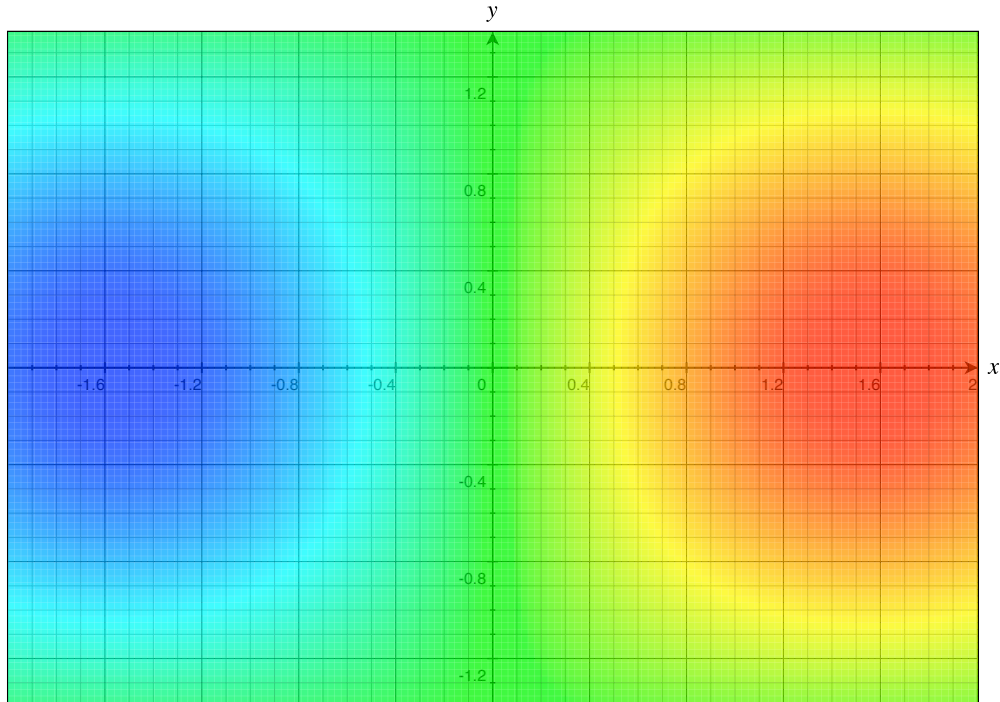


Figure 1: Scalar field plotted for the function $f : x, y \mapsto \sin(x) \cos y$

Definition 1.5. A *vector field* is a function mapping points in space to vector quantities [Brezinski, 2006]. In the case of fluid dynamics, vector fields often model quantities like fluid velocity^[see Figure 2].

Definition 1.6. The *velocity potential* ϕ is a scalar field whose gradient is the velocity vector field of some fluid, mathematically $\mathbf{V} = \nabla \phi$. The quantity is defined for irrotational

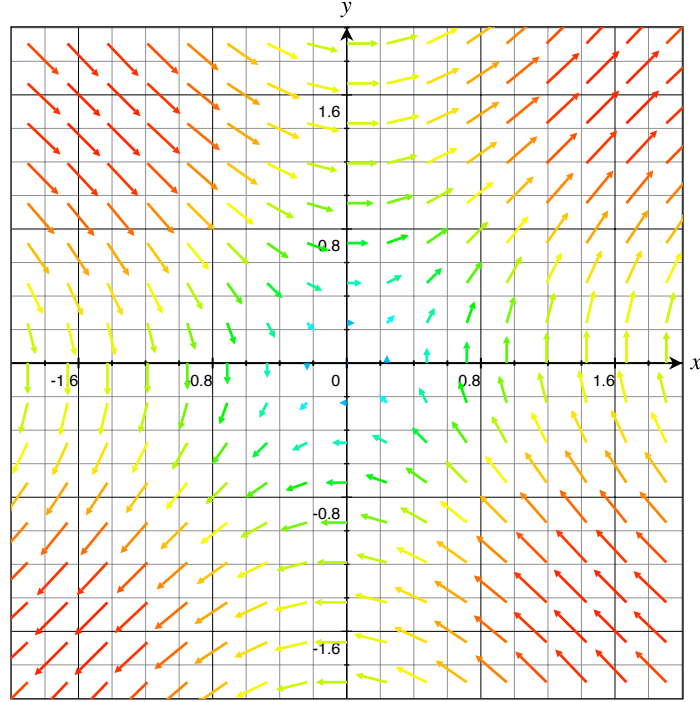


Figure 2: Vector field plotted for the function $\mathbf{F} : x, y \mapsto \begin{pmatrix} \sin y \\ \sin x \end{pmatrix}$

flow (which is a resulting property of the idealisations made in this essay^[see Section 3.2.1]) and is essential to its analysis.

1.2.2 Notation

Vector calculus, like one-variable calculus, has no standardized notation. This essay will employ the following notation:

- ∇ :
 - ∇F : The gradient of some scalar field F .
 - $\nabla \cdot \mathbf{F}$: The divergence of some vector field \mathbf{F} .
 - $\nabla \times \mathbf{F}$: The curl of some vector field \mathbf{F} .
 - $\nabla_{\mathbf{v}} f$: The directional derivative of f in the direction of some vector \mathbf{v}
- Δ : The Laplacian operator
- δ_x : A small change in some variable x , used in place of Δx to avoid confusion with the Laplacian operator.

- $\mathbf{J}_{\mathbf{F}}$: The Jacobian matrix of the function \mathbf{F}
- \mathbf{V}^\top : The transpose of some matrix / vector \mathbf{V}
- $\mathbb{D}_\delta(\langle x, y \rangle)$: The set of the points in an open disk centred at (x, y) with radius δ
- $\hat{i}, \hat{j} \& \hat{k}$: Unit vectors in the positive x, y and z directions respectively.
- $\hat{r} \& \hat{\vartheta}$: Unit vectors in the positive r and ϑ directions respectively.

1.2.3 The mean value theorem

To support the derivations made later in this essay, particularly in the proof of Clairaut's theorem^[see lemma 2.1], fundamental concepts and theorems from single-variable calculus are introduced here, including the Mean Value Theorem and the lemmas it builds upon.

Theorem 1.1 (The extreme value theorem). If a function f is continuous on the finite interval $[a, b]$, then there exists $A, B \in [a, b]$ such that $f(A) \leq f(x) \leq f(B) \forall x \in [a, b]$. Thus, at the points A and B , f has an absolute minimum $m = f(A)$ and an absolute maximum $M = f(B)$.

Theorem 1.2 (Rolle's theorem). If a function f is continuous on the interval $[a, b]$ and differentiable on (a, b) , and $f(a) = f(b)$, then $\exists c \in (a, b)$ such that $f'(c) = 0$.

Proof. Consider two cases:

Case 1: f remains constant over $[a, b]$

If $f(x) = f(a) = f(b) \forall x \in (a, b)$, then $f'(x) = 0$, and the theorem holds trivially.

Case 2: f is not constant over $[a, b]$

If f is not constant over $[a, b]$ and $f(a) = f(b)$, then Theorem 1.1 asserts that there must exist an absolute maximum or minimum that occur at some point $\eta \in (a, b)$. Since f is differentiable over (a, b) , then any point η where an absolute extremum occurs must also be a local extremum. Consider the case where η is a local maxima (the proof for the case of local minima is analogous). Then let the interval $I = (\eta - \delta, \eta + \delta)$ for some $\delta > 0$ such that $\forall X \in I, f(X) \leq f(\eta)$.

Let $h < 0$ be a number sufficiently small such that $\eta + h \in I$. $f(\eta + h) \leq f(\eta) \implies f(\eta + h) - f(\eta) \leq 0$. Thus,

$$\frac{f(\eta + h) - f(\eta)}{h} \geq 0 \because \begin{cases} f(\eta + h) - f(\eta) & \leq 0 \\ h & \leq 0 \end{cases}$$

Taking the left-hand limit as $h \rightarrow 0$,

$$\lim_{h \rightarrow 0^-} \frac{f(\eta + h) - f(\eta)}{h} = f'(\eta)$$

Now let $H > 0$ be a number sufficiently small such that $\eta - H \in I$.

$$\begin{aligned} \frac{f(\eta + H) - f(\eta)}{H} &\leq 0 \because \begin{cases} f(\eta + H) - f(\eta) & \leq 0 \\ H & \geq 0 \end{cases} \\ \lim_{H \rightarrow 0^+} \frac{f(\eta + H) - f(\eta)}{H} &= f'(\eta) \end{aligned}$$

Thus,

$$\begin{aligned} 0 &\geq \lim_{H \rightarrow 0^+} \frac{f(\eta + H) - f(\eta)}{H} = f'(\eta) = \lim_{h \rightarrow 0^-} \frac{f(\eta + h) - f(\eta)}{h} \geq 0 \\ &\therefore f'(\eta) = 0 \end{aligned}$$

Since the same would apply for local minima, then for any local extrema $\eta \in (a, b)$, of which Theorem 1.1 asserts there must exist at least one, $f'(\eta) = 0$. ■

Theorem 1.3 (The mean value theorem). For any function f continuous on the interval $[a, b]$ and differentiable on the interval (a, b) , $\exists c \in (a, b)$ such that

$$f'(c) = \frac{f(b) - f(a)}{b - a} \tag{1}$$

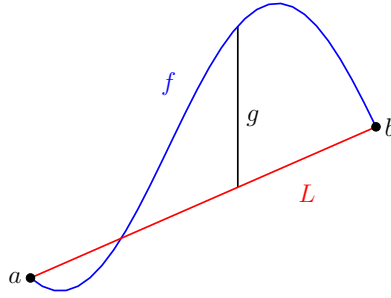
Proof. Consider the region of some function f on the finite interval $[a, b]$ over which f is continuous and differentiable over (a, b) . Let the function L represent the straight line

between the points $(a, f(a))$ and $(b, f(b))$, which is given by the expression:

$$L(x) = f(a) + \frac{f(b) - f(a)}{b - a}(x - a)$$

Now consider the function g , defined as the difference between f and L :

$$g(x) = L(x) - f(x) = f(a) + \frac{f(b) - f(a)}{b - a}(x - a) - f(x)$$



Computing the derivative of g with respect to x gives:

$$g'(x) = \frac{f(b) - f(a)}{b - a} - f'(x)$$

Since $g(a) = g(b) = 0$, Theorem 1.2 asserts that there is at least one point $c \in (a, b)$ such that $g'(c) = 0$. Thus, at c ,

$$\begin{aligned} 0 = g'(c) &= \frac{f(b) - f(a)}{b - a} - f'(c) \\ \implies f'(c) &= \frac{f(b) - f(a)}{b - a} \end{aligned}$$

■

2 Vector calculus

2.1 The fundamentals of vector calculus

Definition 2.1. *Partial derivatives* are a multivariable extension of single-variable derivatives in which all variables save the one being differentiated by are treated as constants [Mor-

timer, 2013]. A formal definition of the partial derivative of some function f with respect to a parameter x_n can be expressed as:

$$\frac{\partial f}{\partial x_n} = \lim_{\delta_x \rightarrow 0} \frac{f(x_1, x_2, \dots, x_n + \delta_x, \dots) - f(x_1, x_2, \dots, x_n, \dots)}{\delta_x} \quad (2)$$

Partial derivatives allow for the analysis of how multi-variable functions such as scalar- or vector fields change with respect to just one spatial dimension. For example, consider the function $f(x, y) = x^2y + \sin(x) \sin y$:

$$\frac{\partial f}{\partial x} = 2xy + \cos(x) \sin y \qquad \frac{\partial f}{\partial y} = x^2 + \sin(x) \cos y$$

n -th order partial derivatives are denoted, similarly to normal calculus, as

$$\frac{\partial^n f}{\partial x^n} \equiv \underbrace{\frac{\partial}{\partial x} \cdots \frac{\partial}{\partial x}}_{n \text{ times}} \frac{\partial f}{\partial x}$$

Definition 2.2. *Mixed partial derivatives* are partial derivatives of a function taken with respect to multiple variables [Garrett, 2015]. This is denoted as

$$\frac{\partial^2 f}{\partial \alpha \partial \beta} \equiv \frac{\partial}{\partial \beta} \frac{\partial f}{\partial \alpha}$$

where both α and β are parameters of f . The order of mixed partial derivatives, provided that both of the first order partial derivatives are continuous, does not matter, as guaranteed by Theorem 2.1 (Clairaut's theorem).

Theorem 2.1 (Clairaut's theorem). Let $f(\alpha, \beta)$ be a function of two parameters α and β . If the mixed partial derivatives $\frac{\partial^2 f}{\partial \alpha \partial \beta}$ and $\frac{\partial^2 f}{\partial \beta \partial \alpha}$ exist and are continuous in the open disk $\mathbb{D}_\delta(\langle \alpha_0, \beta_0 \rangle)$ centred at (α_0, β_0) with radius $\delta > 0$, then

$$\left. \frac{\partial^2 f}{\partial \alpha \partial \beta} \right|_{(\alpha_0, \beta_0)} = \left. \frac{\partial^2 f}{\partial \beta \partial \alpha} \right|_{(\alpha_0, \beta_0)}$$

[Garrett, 2015]

Proof. Let (α_0, β_0) and (α_1, β_1) be points in the domain of f . Consider a rectangular region bound by the points $W(\alpha_0, \beta_0)$, $X(\alpha_1, \beta_0)$, $Y(\alpha_1, \beta_1)$ and $Z(\alpha_0, \beta_1)$. $\frac{\partial f}{\partial \alpha}$ and $\frac{\partial f}{\partial \beta}$ exist in a neighbourhood of this rectangle, and the mixed partial derivatives $\frac{\partial^2 f}{\partial \beta \partial \alpha}$ and $\frac{\partial^2 f}{\partial \alpha \partial \beta}$ exist and are continuous in this neighbourhood. Let Q be such that

$$Q = [f(\alpha_1, \beta_1) - f(\alpha_0, \beta_1)] - [f(\alpha_1, \beta_0) - f(\alpha_0, \beta_0)]$$

According to Theorem 1.3, the mean value theorem (MVT), $\exists \xi_0, \xi_1 \in [\alpha_0, \alpha_1]$ such that

$$\begin{aligned} \left. \frac{\partial f}{\partial \alpha} \right|_{(\xi_0, \beta_0)} &= \frac{f(\alpha_1, \beta_0) - f(\alpha_0, \beta_0)}{\alpha_1 - \alpha_0} \\ \left. \frac{\partial f}{\partial \alpha} \right|_{(\xi_1, \beta_1)} &= \frac{f(\alpha_1, \beta_1) - f(\alpha_0, \beta_1)}{\alpha_1 - \alpha_0} \end{aligned}$$

Thus Q can be expressed as

$$\begin{aligned} Q &= \left(\left. \frac{\partial f}{\partial \alpha} \right|_{(\xi_0, \beta_0)} (\alpha_1 - \alpha_0) \right) - \left(\left. \frac{\partial f}{\partial \alpha} \right|_{(\xi_1, \beta_1)} (\alpha_1 - \alpha_0) \right) \\ &= \left(\left. \frac{\partial f}{\partial \alpha} \right|_{(\xi_0, \beta_0)} - \left. \frac{\partial f}{\partial \alpha} \right|_{(\xi_1, \beta_1)} \right) (\alpha_1 - \alpha_0) \end{aligned}$$

Now let R be the equivalent of Q in the direction of β ,

$$R = [f(\alpha_1, \beta_1) - f(\alpha_1, \beta_0)] - [f(\alpha_0, \beta_1) - f(\alpha_0, \beta_0)]$$

By the MVT $\exists \zeta_0, \zeta_1 \in [\beta_0, \beta_1]$ such that

$$\begin{aligned} \left. \frac{\partial f}{\partial \beta} \right|_{(\alpha_0, \zeta_0)} &= \frac{f(\alpha_0, \beta_1) - f(\alpha_0, \beta_0)}{\beta_1 - \beta_0} \\ \left. \frac{\partial f}{\partial \beta} \right|_{(\alpha_1, \zeta_1)} &= \frac{f(\alpha_1, \beta_1) - f(\alpha_1, \beta_0)}{\beta_1 - \beta_0} \end{aligned}$$

Thus R can be expressed as

$$\begin{aligned} R &= \left(\frac{\partial f}{\partial \beta} \Big|_{(\alpha_0, \zeta_0)} (\beta_1 - \beta_0) \right) - \left(\frac{\partial f}{\partial \beta} \Big|_{(\alpha_1, \zeta_1)} (\beta_1 - \beta_0) \right) \\ &= \left(\frac{\partial f}{\partial \beta} \Big|_{(\alpha_0, \zeta_0)} - \frac{\partial f}{\partial \beta} \Big|_{(\alpha_1, \zeta_1)} \right) (\beta_1 - \beta_0) \end{aligned}$$

Rearranging Q and R ,

$$\begin{aligned} Q &= [f(\alpha_1, \beta_1) - f(\alpha_0, \beta_1)] - [f(\alpha_1, \beta_0) - f(\alpha_0, \beta_0)] \\ &= f(\alpha_1, \beta_1) - f(\alpha_1, \beta_0) - f(\alpha_0, \beta_1) + f(\alpha_0, \beta_0) \\ &= [f(\alpha_1, \beta_1) - f(\alpha_1, \beta_0)] - [f(\alpha_0, \beta_1) - f(\alpha_0, \beta_0)] = R \\ \therefore Q &= R \end{aligned}$$

Thus

$$\begin{aligned} \left(\frac{\partial f}{\partial \alpha} \Big|_{(\xi_0, \beta_0)} - \frac{\partial f}{\partial \alpha} \Big|_{(\xi_1, \beta_1)} \right) (\alpha_1 - \alpha_0) &= \left(\frac{\partial f}{\partial \beta} \Big|_{(\alpha_0, \zeta_0)} - \frac{\partial f}{\partial \beta} \Big|_{(\alpha_1, \zeta_1)} \right) (\beta_1 - \beta_0) \\ \rightsquigarrow \frac{\partial f / \partial \alpha \Big|_{(\xi_0, \beta_0)} - \partial f / \partial \alpha \Big|_{(\xi_1, \beta_1)}}{\beta_1 - \beta_0} &= \frac{\partial f / \partial \beta \Big|_{(\alpha_0, \zeta_0)} - \partial f / \partial \beta \Big|_{(\alpha_1, \zeta_1)}}{\alpha_1 - \alpha_0} \end{aligned} \quad (3)$$

Applying the MVT again $\exists \xi^* \in (\xi_0, \xi_1), \beta^* \in (\beta_0, \beta_1)$ such that

$$\begin{aligned} \frac{\partial^2 f}{\partial \alpha \partial \beta} \Big|_{(\xi^*, \beta^*)} &= \frac{\partial f / \partial \alpha \Big|_{(\xi_1, \beta_1)} - \partial f / \partial \alpha \Big|_{(\xi_0, \beta_0)}}{\beta_1 - \beta_0} \\ \implies - \frac{\partial^2 f}{\partial \alpha \partial \beta} \Big|_{(\xi^*, \beta^*)} &= \frac{\partial f / \partial \alpha \Big|_{(\xi_0, \beta_0)} - \partial f / \partial \alpha \Big|_{(\xi_1, \beta_1)}}{\beta_1 - \beta_0} \end{aligned}$$

Similarly, $\exists \alpha^* \in (\alpha_0, \alpha_1), \zeta^* \in (\zeta_0, \zeta_1)$ such that

$$\begin{aligned} \frac{\partial^2 f}{\partial \beta \partial \alpha} \Big|_{(\alpha^*, \zeta^*)} &= \frac{\partial f / \partial \beta \Big|_{(\alpha_1, \zeta_1)} - \partial f / \partial \beta \Big|_{(\alpha_0, \zeta_0)}}{\alpha_1 - \alpha_0} \\ \implies - \frac{\partial^2 f}{\partial \beta \partial \alpha} \Big|_{(\alpha^*, \zeta^*)} &= \frac{\partial f / \partial \beta \Big|_{(\alpha_0, \zeta_0)} - \partial f / \partial \beta \Big|_{(\alpha_1, \zeta_1)}}{\alpha_1 - \alpha_0} \end{aligned}$$

Substituting back into (3),

$$\begin{aligned} -\frac{\partial^2 f}{\partial \alpha \partial \beta} \Big|_{(\xi^*, \beta^*)} &= -\frac{\partial^2 f}{\partial \beta \partial \alpha} \Big|_{(\alpha^*, \zeta^*)} \\ \Rightarrow \frac{\partial^2 f}{\partial \alpha \partial \beta} \Big|_{(\xi^*, \beta^*)} &= \frac{\partial^2 f}{\partial \beta \partial \alpha} \Big|_{(\alpha^*, \zeta^*)} \end{aligned}$$

Consequently, as $\alpha_1 \rightarrow \alpha_0$ and $\beta_1 \rightarrow \beta_0$, $\xi^* \rightarrow \alpha_0$, $\beta^* \rightarrow \beta_0$, $\alpha^* \rightarrow \alpha_0$ and $\zeta^* \rightarrow \beta_0$. Since the derivatives are continuous,

$$\frac{\partial^2 f}{\partial \beta \partial \alpha} \Big|_{(\alpha_0, \beta_0)} = \frac{\partial^2 f}{\partial \alpha \partial \beta} \Big|_{(\alpha_0, \beta_0)}$$

Because (α_0, β_0) is an arbitrary point in the domain, $\frac{\partial^2 f}{\partial \beta \partial \alpha} = \frac{\partial^2 f}{\partial \alpha \partial \beta}$ at all points in the domain where the mixed partial derivatives are continuous. ■

Definition 2.3. The *nabla* or *gradient* operator ∇ is a vector containing the partial derivatives of a scalar valued function with respect to each of its parameters [Rapp, 2017]. Thus applying the operator is taking the product of the vector ∇ and a scalar valued function f . For some function $f : \mathbb{R}^n \rightarrow \mathbb{R}$, ∇ and ∇f would be given by:

$$\nabla = \begin{bmatrix} \partial/\partial x_1 \\ \partial/\partial x_2 \\ \vdots \\ \partial/\partial x_n \end{bmatrix}, \quad \nabla f = \begin{bmatrix} \partial f/\partial x_1 \\ \partial f/\partial x_2 \\ \vdots \\ \partial f/\partial x_n \end{bmatrix}$$

2.2 Divergence & curl

Definition 2.4. The *divergence* of a vector field \mathbf{F} is a scalar field denoted as $\nabla \cdot \mathbf{F}$, defined as the dot product of the nabla operator (∇) and the vector field. For the function $\mathbf{F} : x, y \mapsto X(x, y)\hat{i} + Y(x, y)\hat{j}$, $\nabla \cdot \mathbf{F}$ would be given by:

$$\nabla \cdot \mathbf{F} = \frac{\partial X}{\partial x} + \frac{\partial Y}{\partial y}$$

Definition 2.5. The *curl* of a vector field \mathbf{F} is a vector field denoted as $\nabla \times \mathbf{F}$, defined as the cross product of the nabla operator and the vector field. Typically, since the cross product is only defined for 3 dimensional spaces, 2 dimensional curl is defined as a scalar

field which for some function $\mathbf{F} : x, y \mapsto X(x, y)\hat{i} + Y(x, y)\hat{j}$ is derived from the coefficient of \hat{k} if the cross product was done as if the z term of both vectors was set to 0. Thus, for such a function \mathbf{F} :

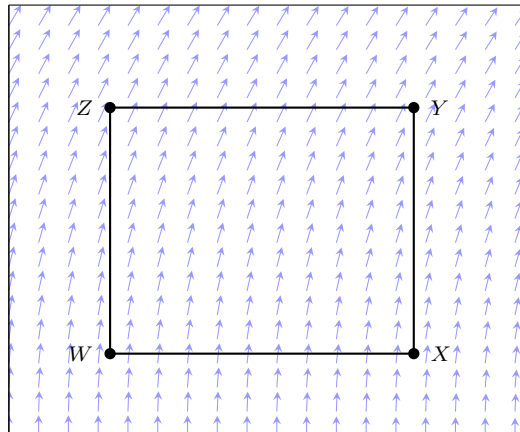
$$\begin{aligned}\nabla \times \mathbf{F} &= \begin{vmatrix} \hat{i} & \hat{j} & \hat{k} \\ \partial/\partial x & \partial/\partial y & 0 \\ X & Y & 0 \end{vmatrix} \\ &= \begin{vmatrix} \partial/\partial y & 0 \\ Y & 0 \end{vmatrix} \hat{i} - \begin{vmatrix} \partial/\partial x & 0 \\ X & 0 \end{vmatrix} \hat{j} + \begin{vmatrix} \partial/\partial x & \partial/\partial y \\ X & Y \end{vmatrix} \hat{k} \\ &= 0\hat{i} - 0\hat{j} + \left(\frac{\partial Y}{\partial x} - \frac{\partial X}{\partial y} \right) \hat{k} = \left(\frac{\partial Y}{\partial x} - \frac{\partial X}{\partial y} \right) \hat{k} \\ &\rightsquigarrow \nabla \times \mathbf{F} = \frac{\partial Y}{\partial x} - \frac{\partial X}{\partial y}\end{aligned}$$

3 Governing equations for ideal fluid flow

3.1 The continuity equation for incompressible flow

The idealised fluid in this essay is incompressible. This means that if ρ represents the density of the fluid, then as per definition 1.3, ρ must remain constant over time at every point in the domain of the vector field representing the fluid flow. This means that for any arbitrary closed volume within the fluid, the net mass flow rate across its boundaries must be zero.

Let the velocity vector field of the fluid be $\mathbf{V} : x, y \mapsto X(x, y)\hat{i} + Y(x, y)\hat{j}$. To quantify the mass flow, consider an arbitrary infinitesimal rectangular volume within the fluid, with vertices $W(\alpha_0, \beta_0)$, $X(\alpha_1, \beta_0)$, $Y(\alpha_1, \beta_1)$, and $Z(\alpha_0, \beta_1)$, as depicted below. Let $\bar{\alpha} = \frac{\alpha_0 + \alpha_1}{2}$ and $\bar{\beta} = \frac{\beta_0 + \beta_1}{2}$. Assume \mathbf{V} is continuous and differentiable over this region.



The mass flow rate (\dot{m} , mass per unit time) across a given surface is defined as the flux of mass, which is computed as the product of density, the velocity component normal to the surface, and the area of the surface. Thus, along the x axis (in the direction of \hat{i}) the mass flow rate into WZ is given as,

$$\dot{m}_{\rightarrow WZ} = \rho X(\alpha_0, \bar{\beta})(\beta_1 - \beta_0)$$

and similarly the mass flow rate out of the opposite side XY is given as:

$$\dot{m}_{XY \rightarrow} = \rho X(\alpha_1, \bar{\beta})(\beta_1 - \beta_0)$$

Thus, the net mass flow rate out of the rectangular region along the x axis is:

$$\begin{aligned} \dot{m}_i &= \dot{m}_{XY \rightarrow} - \dot{m}_{\rightarrow WZ} \\ &= \rho X(\alpha_1, \bar{\beta})(\beta_1 - \beta_0) - \rho X(\alpha_0, \bar{\beta})(\beta_1 - \beta_0) \end{aligned}$$

Which, factoring out $\rho(\beta_1 - \beta_0)$, leads to:

$$\dot{m}_i = \rho(\beta_1 - \beta_0) [X(\alpha_1, \bar{\beta}) - X(\alpha_0, \bar{\beta})]$$

Analogously, across the y axis, the net mass flow rate out of the rectangular region between sides WX and ZY is given by the expression:

$$\dot{m}_j = \rho(\alpha_1 - \alpha_0) [Y(\bar{\alpha}, \beta_1) - Y(\bar{\alpha}, \beta_0)]$$

Therefore, the net mass flow rate out of the rectangular region, which must be equal to 0 for the fluid to be incompressible, is given by:

$$\begin{aligned} \dot{m} &= \dot{m}_i + \dot{m}_j \\ &= \rho(\beta_1 - \beta_0) [X(\alpha_1, \bar{\beta}) - X(\alpha_0, \bar{\beta})] + \rho(\alpha_1 - \alpha_0) [Y(\bar{\alpha}, \beta_1) - Y(\bar{\alpha}, \beta_0)] = 0 \end{aligned}$$

Dividing through by ρ , $(\alpha_1 - \alpha_0)$ and $(\beta_1 - \beta_0)$:

$$\frac{X(\alpha_1, \bar{\beta}) - X(\alpha_0, \bar{\beta})}{\alpha_1 - \alpha_0} + \frac{Y(\bar{\alpha}, \beta_1) - Y(\bar{\alpha}, \beta_0)}{\beta_1 - \beta_0} = 0$$

Now consider the limit as $\alpha_1 \rightarrow \alpha_0$ and $\beta_1 \rightarrow \beta_0$, the difference $\delta_\alpha = \alpha_1 - \alpha_0 \rightarrow 0$ and $\delta_\beta = \beta_1 - \beta_0 \rightarrow 0$.

$$\begin{aligned} \lim_{\delta_\alpha \rightarrow 0} \frac{X(\alpha_0 + \delta_\alpha, \bar{\beta}) - X(\alpha_0, \bar{\beta})}{\delta_\alpha} &= \left. \frac{\partial X}{\partial x} \right|_{(\alpha_0, \bar{\beta})} \\ \lim_{\delta_\beta \rightarrow 0} \frac{Y(\bar{\alpha}, \beta_0 + \delta_\beta) - Y(\bar{\alpha}, \beta_0)}{\delta_\beta} &= \left. \frac{\partial Y}{\partial y} \right|_{(\bar{\alpha}, \beta_0)} \end{aligned}$$

Furthermore, as $\alpha_1 \rightarrow \alpha_0$ and $\beta_1 \rightarrow \beta_0$, $\bar{\alpha} \rightarrow \alpha_0$ and $\bar{\beta} \rightarrow \beta_0$, consequently

$$\dot{m} = \left. \frac{\partial X}{\partial x} \right|_{(\alpha_0, \beta_0)} + \left. \frac{\partial Y}{\partial y} \right|_{(\alpha_0, \beta_0)} = \nabla \cdot \mathbf{V} \Big|_{(\alpha_0, \beta_0)} = 0$$

Because (α_0, β_0) is any point in the domain of \mathbf{V} where the function is differentiable, the expression can be generalised as

$$\nabla \cdot \mathbf{V} = 0$$

This equation is known as the continuity equation for incompressible fluids [Pham et al., 2014] and will underpin following derivations made in this essay.

3.2 Irrotational flow

As mentioned in definition 1.6, one resulting property of the idealisations (steady, inviscid and incompressible flow) made in this essay is irrotational flow. If flow is rotational, then there exists points at which $\nabla \times \mathbf{F} \neq 0$. In other words, if one were to imagine a water wheel at some point in the fluid, and it spins, then the flow is rotational, and vice versa for irrotational flow. However, flow being irrotational does not imply that it cannot curve, for

example $\nabla \times \mathbf{F} = 0$ in cases such as:

$$\begin{aligned}\mathbf{F} : x, y &\mapsto X(x, y)\hat{i} + Y(x, y)\hat{j} \quad \forall (x, y) \neq (0, 0) \\ X : x, y &\mapsto -\frac{y}{x^2 + y^2}, \quad Y : x, y \mapsto \frac{x}{x^2 + y^2}\end{aligned}$$

Applying the quotient rule to compute the derivatives for both X and Y gives:

$$\begin{aligned}\frac{\partial X}{\partial y} &= -\frac{\left(\frac{\partial}{\partial y}y\right)(x^2 + y^2) - y\frac{\partial}{\partial y}(x^2 + y^2)}{(x^2 + y^2)^2} \\ &= -\frac{x^2 + y^2 - y(2y)}{(x^2 + y^2)^2} \\ &= -\frac{x^2 - y^2}{(x^2 + y^2)^2}\end{aligned}$$

$$\begin{aligned}\frac{\partial Y}{\partial x} &= \frac{\left(\frac{\partial}{\partial x}x\right)(x^2 + y^2) - x\frac{\partial}{\partial x}(x^2 + y^2)}{(x^2 + y^2)^2} \\ &= \frac{x^2 + y^2 - x(2x)}{(x^2 + y^2)^2} \\ &= \frac{-x^2 + y^2}{(x^2 + y^2)^2} = -\frac{x^2 - y^2}{(x^2 + y^2)^2}\end{aligned}$$

Thus,

$$\begin{aligned}\frac{\partial X}{\partial y} = \frac{\partial Y}{\partial x} &\implies \frac{\partial Y}{\partial x} - \frac{\partial X}{\partial y} = 0 \\ &\therefore \nabla \times \mathbf{F} = 0\end{aligned}$$

Plotting the vector field for \mathbf{F} reveals circulation around the origin, suggesting rotational flow, but which, with a curl of 0 (everywhere except for the origin, where \mathbf{F} is undefined), is irrotational^[see figure 3].

3.2.1 Kelvin's circulation theorem

Kelvin's circulation theorem is crucial in the modelling of ideal fluids. While a full proof of the theorem is outside the scope of this essay for the purposes of keeping focused, it provides

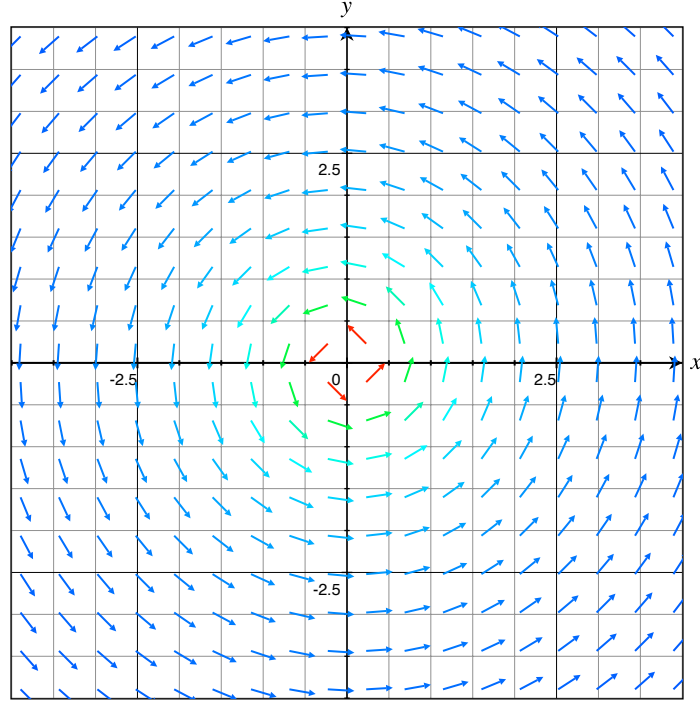


Figure 3: The function $\mathbf{F} : x, y \mapsto \begin{pmatrix} -y(x^2 + y^2)^{-1} \\ x(x^2 + y^2)^{-1} \end{pmatrix}$ is irrotational despite curving

the guarantee of irrotational flow which underpins the rest of this essay.

The theorem states that for an ideal (crucially inviscid) fluid, the circulation around a closed loop moving with the fluid remains constant over time. Therefore, implying that if a fluid begins its flow irrotational, it remains irrotational. Irrotational flow enables the use of a velocity potential^[see Section 3.3], greatly simplifying the further analysis of the fluid flow.

Theorem 3.1 (Kelvin’s circulation theorem). The rate of change of the circulation (Γ) of an ideal and inviscid fluid with respect to time around a closed contour moving with the flow of the fluid (in a velocity field \mathbf{V}) is zero. Mathematically,

$$\overbrace{\frac{\partial \Gamma}{\partial t} + \mathbf{V} \cdot \nabla \Gamma}^{\text{Commonly denoted as } \frac{D\Gamma}{Dt}} = 0$$

3.3 The velocity potential

As established in Section 3.2.1 through Theorem 3.1 (Kelvin's circulation theorem), for an incompressible and inviscid fluid initially irrotational, the flow remains irrotational, meaning the velocity field \mathbf{V} modelling the flow satisfies $\nabla \times \mathbf{V} = 0$. This condition leads to the existence of a scalar-valued function, known as the velocity potential, whose gradient is equal to the vector field. To demonstrate this, consider the identity $\nabla \times \nabla \phi$, where ϕ is a function of x and y and is scalar valued. $\nabla \phi$ is defined as:

$$\nabla \phi = \begin{bmatrix} \partial \phi / \partial x \\ \partial \phi / \partial y \end{bmatrix}$$

Taking the curl of this expression and applying Theorem 2.1 (Clairaut's theorem) gives

$$\begin{aligned} \nabla \times \nabla \phi &= \frac{\partial^2 \phi}{\partial y \partial x} - \frac{\partial^2 \phi}{\partial x \partial y} \\ &= \frac{\partial^2 \phi}{\partial y \partial x} - \frac{\partial^2 \phi}{\partial y \partial x} = 0 \end{aligned}$$

Thus, by the definition of irrotational flow,

$$\begin{aligned} \nabla \times \mathbf{V} &= 0 = \nabla \times \nabla \phi \\ \therefore \mathbf{V} &= \nabla \phi \end{aligned}$$

Here, ϕ is known as the velocity potential of \mathbf{V} . The velocity potential makes derivations made in this essay easier due its scalar valued nature.

3.4 The multivariable chain rule

3.4.1 The Jacobian matrix

Definition 3.1. The *Jacobian matrix* of some function \mathbf{F} , denoted $\mathbf{J}_{\mathbf{F}}$, is defined as the matrix containing all the first order partial derivatives of the function. For the function

$$\mathbf{F} : x_1, x_2, \dots, x_n \mapsto \begin{bmatrix} f_1(x_1, x_2, \dots, x_n) \\ f_2(x_1, x_2, \dots, x_n) \\ \vdots \\ f_m(x_1, x_2, \dots, x_n) \end{bmatrix}$$

the Jacobian of \mathbf{F} would be given as

$$\mathbf{J}_{\mathbf{F}} = \begin{bmatrix} \nabla^\top f_1 \\ \nabla^\top f_2 \\ \vdots \\ \nabla^\top f_m \end{bmatrix} = \begin{bmatrix} \partial f_1 / \partial x_1 & \partial f_1 / \partial x_2 & \cdots & \partial f_1 / \partial x_n \\ \partial f_2 / \partial x_1 & \partial f_2 / \partial x_2 & \cdots & \partial f_2 / \partial x_n \\ \vdots & \vdots & \ddots & \vdots \\ \partial f_m / \partial x_1 & \partial f_m / \partial x_2 & \cdots & \partial f_m / \partial x_n \end{bmatrix} \in \mathbb{R}^{m \times n}$$

The Jacobian matrix, and its determinant, has many uses within vector calculus, but for the purposes of this essay, the Jacobian chain rule is a fundamental result which allows for the calculation of the derivatives of a composite functions. The following result and proof underpins the proof for the more generalised version in Section 3.4.2.

Lemma 3.2 (The Jacobian chain rule). Let the function $\mathbf{F} : \mathbb{R}^m \rightarrow \mathbb{R}^n$ and the function $\mathbf{G} : \mathbb{R}^l \rightarrow \mathbb{R}^m$. The Jacobian of the composite $\mathbf{F} \circ \mathbf{G}$ evaluated at some point \mathbf{p} can be expressed as

$$\mathbf{J}_{\mathbf{F} \circ \mathbf{G}}(\mathbf{p}) = [\mathbf{J}_{\mathbf{F}} \circ \mathbf{G}(\mathbf{p})] \mathbf{J}_{\mathbf{G}}(\mathbf{p})$$

Proof. A function $\mathbf{F} : \mathbb{R}^m \rightarrow \mathbb{R}^n$ is considered differentiable at some point \mathbf{a} if there exists a linear map such (represented by its Jacobian matrix $\mathbf{J}_{\mathbf{F}}(\mathbf{a})$) that for an infinitesimal vector \mathbf{h}

$$\mathbf{F}(\mathbf{a} + \mathbf{h}) - \mathbf{F}(\mathbf{a}) = \mathbf{J}_{\mathbf{F}}(\mathbf{a})\mathbf{h} + \epsilon(\mathbf{h})$$

where $\epsilon(\mathbf{h})$ is an error term such that it vanishes faster than the magnitude of \mathbf{h} :

$$\lim_{\mathbf{h} \rightarrow \mathbf{0}} \frac{\|\epsilon(\mathbf{h})\|}{\|\mathbf{h}\|} = \mathbf{0} \quad (4)$$

Consider the function $\mathbf{Z} : \mathbf{t} \mapsto (\mathbf{F} \circ \mathbf{G})(\mathbf{t})$, where $\mathbf{G} : \mathbb{R}^l \rightarrow \mathbb{R}^m$ and $\mathbf{F} : \mathbb{R}^m \rightarrow \mathbb{R}^n$. For some small change in \mathbf{t} , called $\delta_{\mathbf{t}}$, the change in the inner function \mathbf{G} can be expressed using its differentiability at \mathbf{t}

$$\delta_{\mathbf{G}} = \mathbf{G}(\mathbf{t} + \delta_{\mathbf{t}}) - \mathbf{G}(\mathbf{t}) = \mathbf{J}_{\mathbf{G}}(\mathbf{t})\delta_{\mathbf{t}} + \epsilon_{\mathbf{G}}(\delta_{\mathbf{t}}) \quad (5)$$

The error term $\epsilon_{\mathbf{G}}(\delta_{\mathbf{t}})$ satisfies (4). Then consider the change of the outer function \mathbf{F} , also expressed using its differentiability at the point $\mathbf{u} = \mathbf{G}(\mathbf{t})$,

$$\delta_{\mathbf{F}} = \mathbf{F}(\mathbf{u} + \delta_{\mathbf{G}}) - \mathbf{F}(\mathbf{u}) = \mathbf{J}_{\mathbf{F}}(\mathbf{u})\delta_{\mathbf{G}} + \epsilon_{\mathbf{F}}(\delta_{\mathbf{G}})$$

Which, since $\mathbf{Z} = (\mathbf{F} \circ \mathbf{G})(\mathbf{t})$, means that

$$\delta_{\mathbf{Z}} = \delta_{\mathbf{F}} = \mathbf{F}(\mathbf{u} + \delta_{\mathbf{G}}) - \mathbf{F}(\mathbf{u}) = \mathbf{J}_{\mathbf{F}}(\mathbf{u})\delta_{\mathbf{G}} + \epsilon_{\mathbf{F}}(\delta_{\mathbf{G}}) \quad (6)$$

Let the error term $\epsilon_{\mathbf{F}}(\delta_{\mathbf{G}})$ also satisfy (4). Then, substituting (5) into (6),

$$\begin{aligned} \delta_{\mathbf{Z}} &= \mathbf{J}_{\mathbf{F}}(\mathbf{u}) [\mathbf{J}_{\mathbf{G}}(\mathbf{t})\delta_{\mathbf{t}} + \epsilon_{\mathbf{G}}(\delta_{\mathbf{t}})] + \epsilon_{\mathbf{F}}(\mathbf{J}_{\mathbf{G}}(\mathbf{t})\delta_{\mathbf{t}} + \epsilon_{\mathbf{G}}(\delta_{\mathbf{t}})) \\ &= \mathbf{J}_{\mathbf{F}}(\mathbf{u})\mathbf{J}_{\mathbf{G}}(\mathbf{t})\delta_{\mathbf{t}} + \mathbf{J}_{\mathbf{F}}(\mathbf{u})\epsilon_{\mathbf{G}}(\delta_{\mathbf{t}}) + \epsilon_{\mathbf{F}}(\mathbf{J}_{\mathbf{G}}(\mathbf{t})\delta_{\mathbf{t}} + \epsilon_{\mathbf{G}}(\delta_{\mathbf{t}})) \end{aligned}$$

Dividing through by $\|\delta_{\mathbf{t}}\|$,

$$\frac{\delta_{\mathbf{Z}}}{\|\delta_{\mathbf{t}}\|} = \mathbf{J}_{\mathbf{F}}(\mathbf{u})\mathbf{J}_{\mathbf{G}}(\mathbf{t}) \overbrace{\frac{\delta_{\mathbf{t}}}{\|\delta_{\mathbf{t}}\|}}^{\text{Henceforth } \hat{\delta}_{\mathbf{t}}}} + \mathbf{J}_{\mathbf{F}}(\mathbf{u}) \frac{\epsilon_{\mathbf{G}}(\delta_{\mathbf{t}})}{\|\delta_{\mathbf{t}}\|} + \frac{\epsilon_{\mathbf{F}}(\mathbf{J}_{\mathbf{G}}(\mathbf{t})\delta_{\mathbf{t}} + \epsilon_{\mathbf{G}}(\delta_{\mathbf{t}}))}{\|\delta_{\mathbf{t}}\|}$$

Considering the value of the error terms as $\delta_{\mathbf{t}} \rightarrow \mathbf{0}$,

$$\lim_{\delta_{\mathbf{t}} \rightarrow \mathbf{0}} \mathbf{J}_{\mathbf{F}}(\mathbf{u}) \frac{\epsilon_{\mathbf{G}}(\delta_{\mathbf{t}})}{\|\delta_{\mathbf{t}}\|} + \frac{\epsilon_{\mathbf{F}}(\mathbf{J}_{\mathbf{G}}(\mathbf{t})\delta_{\mathbf{t}} + \epsilon_{\mathbf{G}}(\delta_{\mathbf{t}}))}{\|\delta_{\mathbf{t}}\|}$$

For any matrix \mathbf{M} and vector \mathbf{v} , the magnitude of their product $\|\mathbf{M}\mathbf{v}\| \leq \|\mathbf{M}\|\|\mathbf{v}\|$, thus for

the first term,

$$\begin{aligned}\lim_{\delta_{\mathbf{t}} \rightarrow \mathbf{0}} \left\| \mathbf{J}_{\mathbf{F}}(\mathbf{u}) \frac{\epsilon_{\mathbf{G}}(\delta_{\mathbf{t}})}{\|\delta_{\mathbf{t}}\|} \right\| &\leq \lim_{\delta_{\mathbf{t}} \rightarrow \mathbf{0}} \|\mathbf{J}_{\mathbf{F}}(\mathbf{u})\| \left\| \frac{\epsilon_{\mathbf{G}}(\delta_{\mathbf{t}})}{\|\delta_{\mathbf{t}}\|} \right\| \\ &= \lim_{\delta_{\mathbf{t}} \rightarrow \mathbf{0}} \|\mathbf{J}_{\mathbf{F}}(\mathbf{u})\| \frac{\|\epsilon_{\mathbf{G}}(\delta_{\mathbf{t}})\|}{\|\delta_{\mathbf{t}}\|}\end{aligned}$$

By the definition of the definition of the error term $\lim_{\delta_{\mathbf{t}} \rightarrow \mathbf{0}} \frac{\|\epsilon_{\mathbf{G}}(\delta_{\mathbf{t}})\|}{\|\delta_{\mathbf{t}}\|} = 0$, therefore

$$\begin{aligned}\lim_{\delta_{\mathbf{t}} \rightarrow \mathbf{0}} \left\| \mathbf{J}_{\mathbf{F}}(\mathbf{u}) \frac{\epsilon_{\mathbf{G}}(\delta_{\mathbf{t}})}{\|\delta_{\mathbf{t}}\|} \right\| &\leq \lim_{\delta_{\mathbf{t}} \rightarrow \mathbf{0}} \|\mathbf{J}_{\mathbf{F}}(\mathbf{u})\| \frac{\|\epsilon_{\mathbf{G}}(\delta_{\mathbf{t}})\|}{\|\delta_{\mathbf{t}}\|} \\ \implies \lim_{\delta_{\mathbf{t}} \rightarrow \mathbf{0}} \left\| \mathbf{J}_{\mathbf{F}}(\mathbf{u}) \frac{\epsilon_{\mathbf{G}}(\delta_{\mathbf{t}})}{\|\delta_{\mathbf{t}}\|} \right\| &\leq 0 \\ \therefore \lim_{\delta_{\mathbf{t}} \rightarrow \mathbf{0}} \left\| \mathbf{J}_{\mathbf{F}}(\mathbf{u}) \frac{\epsilon_{\mathbf{G}}(\delta_{\mathbf{t}})}{\|\delta_{\mathbf{t}}\|} \right\| = 0 &\implies \lim_{\delta_{\mathbf{t}} \rightarrow \mathbf{0}} \mathbf{J}_{\mathbf{F}}(\mathbf{u}) \frac{\epsilon_{\mathbf{G}}(\delta_{\mathbf{t}})}{\|\delta_{\mathbf{t}}\|} = \mathbf{0}\end{aligned}$$

The second error term can be shown to also be equal to $\mathbf{0}$ through rearrangment,

$$\begin{aligned}\lim_{\delta_{\mathbf{t}} \rightarrow \mathbf{0}} \frac{\epsilon_{\mathbf{F}}(\mathbf{J}_{\mathbf{G}}(\mathbf{t})\delta_{\mathbf{t}} + \epsilon_{\mathbf{G}}(\delta_{\mathbf{t}}))}{\|\delta_{\mathbf{t}}\|} &= \lim_{\delta_{\mathbf{t}} \rightarrow \mathbf{0}} \frac{\epsilon_{\mathbf{F}}(\mathbf{J}_{\mathbf{G}}(\mathbf{t})\delta_{\mathbf{t}} + \epsilon_{\mathbf{G}}(\delta_{\mathbf{t}}))}{\|\mathbf{J}_{\mathbf{G}}(\mathbf{t})\delta_{\mathbf{t}} + \epsilon_{\mathbf{G}}(\delta_{\mathbf{t}})\|} \frac{\|\mathbf{J}_{\mathbf{G}}(\mathbf{t})\delta_{\mathbf{t}} + \epsilon_{\mathbf{G}}(\delta_{\mathbf{t}})\|}{\|\delta_{\mathbf{t}}\|} \\ &= \lim_{\delta_{\mathbf{t}} \rightarrow \mathbf{0}} \frac{\epsilon_{\mathbf{F}}(\mathbf{J}_{\mathbf{G}}(\mathbf{t})\delta_{\mathbf{t}} + \epsilon_{\mathbf{G}}(\delta_{\mathbf{t}}))}{\|\mathbf{J}_{\mathbf{G}}(\mathbf{t})\delta_{\mathbf{t}} + \epsilon_{\mathbf{G}}(\delta_{\mathbf{t}})\|} \lim_{\delta_{\mathbf{t}} \rightarrow \mathbf{0}} \frac{\|\mathbf{J}_{\mathbf{G}}(\mathbf{t})\delta_{\mathbf{t}} + \epsilon_{\mathbf{G}}(\delta_{\mathbf{t}})\|}{\|\delta_{\mathbf{t}}\|} \\ &= \mathbf{0} \lim_{\delta_{\mathbf{t}} \rightarrow \mathbf{0}} \frac{\|\mathbf{J}_{\mathbf{G}}(\mathbf{t})\delta_{\mathbf{t}} + \epsilon_{\mathbf{G}}(\delta_{\mathbf{t}})\|}{\|\delta_{\mathbf{t}}\|} \\ \lim_{\delta_{\mathbf{t}} \rightarrow \mathbf{0}} \frac{\|\mathbf{J}_{\mathbf{G}}(\mathbf{t})\delta_{\mathbf{t}} + \epsilon_{\mathbf{G}}(\delta_{\mathbf{t}})\|}{\|\delta_{\mathbf{t}}\|} &\leq \lim_{\delta_{\mathbf{t}} \rightarrow \mathbf{0}} \frac{\|\mathbf{J}_{\mathbf{G}}(\mathbf{t})\delta_{\mathbf{t}}\|}{\|\delta_{\mathbf{t}}\|} + \frac{\|\epsilon_{\mathbf{G}}(\delta_{\mathbf{t}})\|}{\|\delta_{\mathbf{t}}\|} \\ &= \lim_{\delta_{\mathbf{t}} \rightarrow \mathbf{0}} \frac{\|\mathbf{J}_{\mathbf{G}}(\mathbf{t})\delta_{\mathbf{t}}\|}{\|\delta_{\mathbf{t}}\|} + \mathbf{0} \leq \lim_{\delta_{\mathbf{t}} \rightarrow \mathbf{0}} \frac{\|\mathbf{J}_{\mathbf{G}}(\mathbf{t})\| \|\delta_{\mathbf{t}}\|}{\|\delta_{\mathbf{t}}\|} + \mathbf{0} \\ &= \|\mathbf{J}_{\mathbf{G}}(\mathbf{t})\| \\ \therefore \lim_{\delta_{\mathbf{t}} \rightarrow \mathbf{0}} \frac{\epsilon_{\mathbf{F}}(\mathbf{J}_{\mathbf{G}}(\mathbf{t})\delta_{\mathbf{t}} + \epsilon_{\mathbf{G}}(\delta_{\mathbf{t}}))}{\|\delta_{\mathbf{t}}\|} &\leq \mathbf{0} \|\mathbf{J}_{\mathbf{G}}(\mathbf{t})\| = \mathbf{0} \forall \|\mathbf{J}_{\mathbf{G}}(\mathbf{t})\| \\ \therefore \lim_{\delta_{\mathbf{t}} \rightarrow \mathbf{0}} \frac{\epsilon_{\mathbf{F}}(\mathbf{J}_{\mathbf{G}}(\mathbf{t})\delta_{\mathbf{t}} + \epsilon_{\mathbf{G}}(\delta_{\mathbf{t}}))}{\|\delta_{\mathbf{t}}\|} &= \mathbf{0}\end{aligned}$$

Thus, both of the error terms approach $\mathbf{0}$, meaning that

$$\begin{aligned}
\lim_{\delta_{\mathbf{t}} \rightarrow \mathbf{0}} \frac{\delta_{\mathbf{Z}}}{\|\delta_{\mathbf{t}}\|} &= \lim_{\delta_{\mathbf{t}} \rightarrow \mathbf{0}} \mathbf{J}_{\mathbf{F}}(\mathbf{u})\mathbf{J}_{\mathbf{G}}(\mathbf{t})\hat{\delta}_{\mathbf{t}} + \mathbf{J}_{\mathbf{F}}(\mathbf{u})\frac{\epsilon_{\mathbf{G}}(\delta_{\mathbf{t}})}{\|\delta_{\mathbf{t}}\|} + \frac{\epsilon_{\mathbf{F}}(\mathbf{J}_{\mathbf{G}}(\mathbf{t})\delta_{\mathbf{t}} + \epsilon_{\mathbf{G}}(\delta_{\mathbf{t}}))}{\|\delta_{\mathbf{t}}\|} \\
&= \lim_{\delta_{\mathbf{t}} \rightarrow \mathbf{0}} \mathbf{J}_{\mathbf{F}}(\mathbf{u})\mathbf{J}_{\mathbf{G}}(\mathbf{t})\hat{\delta}_{\mathbf{t}} + \mathbf{0} + \mathbf{0} \\
&= \mathbf{J}_{\mathbf{F}}(\mathbf{G}(\mathbf{t}))\mathbf{J}_{\mathbf{G}}(\mathbf{t})\hat{\delta}_{\mathbf{t}} = [\mathbf{J}_{\mathbf{F}} \circ \mathbf{G}(\mathbf{t})] \mathbf{J}_{\mathbf{G}}(\mathbf{t})\hat{\delta}_{\mathbf{t}}
\end{aligned} \tag{7}$$

Using the definition of differentiability again for \mathbf{Z} at \mathbf{t} ,

$$\begin{aligned}
\delta_{\mathbf{Z}} &= \mathbf{Z}(\mathbf{t} + \delta_{\mathbf{t}}) - \mathbf{Z}(\mathbf{t}) = \mathbf{J}_{\mathbf{Z}}(\mathbf{t})\delta_{\mathbf{t}} + \epsilon_{\mathbf{Z}}(\delta_{\mathbf{t}}) \\
\implies \lim_{\delta_{\mathbf{t}} \rightarrow \mathbf{0}} \frac{\delta_{\mathbf{Z}}}{\|\delta_{\mathbf{t}}\|} &= \mathbf{J}_{\mathbf{Z}}(\mathbf{t})\hat{\delta}_{\mathbf{t}}
\end{aligned} \tag{8}$$

Therefore, substituting (8) into (7),

$$\begin{aligned}
\mathbf{J}_{\mathbf{Z}}(\mathbf{t})\hat{\delta}_{\mathbf{t}} &= [\mathbf{J}_{\mathbf{F}} \circ \mathbf{G}(\mathbf{t})] \mathbf{J}_{\mathbf{G}}(\mathbf{t})\hat{\delta}_{\mathbf{t}} \\
\therefore \mathbf{J}_{\mathbf{F} \circ \mathbf{G}}(\mathbf{t})\hat{\delta}_{\mathbf{t}} &= [\mathbf{J}_{\mathbf{F}} \circ \mathbf{G}(\mathbf{t})] \mathbf{J}_{\mathbf{G}}(\mathbf{t})\hat{\delta}_{\mathbf{t}}
\end{aligned}$$

Now let $\mathbf{A} = \mathbf{J}_{\mathbf{F} \circ \mathbf{G}}(\mathbf{t})$ and $\mathbf{B} = [\mathbf{J}_{\mathbf{F}} \circ \mathbf{G}(\mathbf{t})] \mathbf{J}_{\mathbf{G}}(\mathbf{t})$,

$$\begin{aligned}
\mathbf{A}\hat{\delta}_{\mathbf{t}} &= \mathbf{B}\hat{\delta}_{\mathbf{t}} \implies \mathbf{0} = \mathbf{A}\hat{\delta}_{\mathbf{t}} - \mathbf{B}\hat{\delta}_{\mathbf{t}} \\
&= (\mathbf{A} - \mathbf{B}) \hat{\delta}_{\mathbf{t}}
\end{aligned}$$

Since this equality holds for all non-zero vectors $\hat{\delta}_{\mathbf{t}}$, the matrix mapping the unit vector $\hat{\delta}_{\mathbf{t}}$ must be the zero matrix, consequently,

$$\begin{aligned}
\mathbf{A} - \mathbf{B} &= \mathbf{0} \\
\implies \mathbf{A} &= \mathbf{B} \\
\therefore \mathbf{J}_{\mathbf{F} \circ \mathbf{G}}(\mathbf{t}) &= [\mathbf{J}_{\mathbf{F}} \circ \mathbf{G}(\mathbf{t})] \mathbf{J}_{\mathbf{G}}(\mathbf{t})
\end{aligned}$$

■

3.4.2 The multivariable chain rule

The more generalised form of the multivariable chain rule can now be proven using the Jacobian matrix version of the chain rule proved in Lemma 3.2.

Lemma 3.3 (The multivariable chain rule). Let $X : \mathbb{R}^2 \rightarrow \mathbb{R}$ and $Y : \mathbb{R}^2 \rightarrow \mathbb{R}$ be functions of some parameters α, β . Then let $Z : \mathbb{R}^2 \rightarrow \mathbb{R}$ be a function of the parameters x and y . The partial derivatives of the composite function $z : \alpha, \beta \mapsto Z(X(\alpha, \beta), Y(\alpha, \beta))$ are given by:

$$\begin{aligned}\frac{\partial z}{\partial \alpha} &= \frac{\partial Z}{\partial x} \frac{\partial X}{\partial \alpha} + \frac{\partial Z}{\partial y} \frac{\partial Y}{\partial \alpha} \\ \frac{\partial z}{\partial \beta} &= \frac{\partial Z}{\partial x} \frac{\partial X}{\partial \beta} + \frac{\partial Z}{\partial y} \frac{\partial Y}{\partial \beta}\end{aligned}$$

Proof. Let $\mathbf{G} : \mathbb{R}^l \rightarrow \mathbb{R}^m$ and $\mathbf{F} : \mathbb{R}^m \rightarrow \mathbb{R}^n$. Suppose the output functions and parameters of \mathbf{F} are called f_1, f_2, \dots, f_n and x_1, x_2, \dots, x_m respectively, and the output functions and parameters of \mathbf{G} are called g_1, g_2, \dots, g_m and y_1, y_2, \dots, y_l respectively. The Jacobian matrix of the composite function $\mathbf{F} \circ \mathbf{G}$ is defined as:

$$\mathbf{J}_{\mathbf{F} \circ \mathbf{G}} = \begin{bmatrix} \partial(f_1 \circ \mathbf{G})/\partial y_1 & \partial(f_1 \circ \mathbf{G})/\partial y_2 & \cdots & \partial(f_1 \circ \mathbf{G})/\partial y_l \\ \partial(f_2 \circ \mathbf{G})/\partial y_1 & \partial(f_2 \circ \mathbf{G})/\partial y_2 & \cdots & \partial(f_2 \circ \mathbf{G})/\partial y_l \\ \vdots & \vdots & \ddots & \vdots \\ \partial(f_n \circ \mathbf{G})/\partial y_1 & \partial(f_n \circ \mathbf{G})/\partial y_2 & \cdots & \partial(f_n \circ \mathbf{G})/\partial y_l \end{bmatrix}$$

Thus the element at indices i, j of the Jacobian matrix $\mathbf{J}_{\mathbf{F} \circ \mathbf{G}}$ is given by:

$$(\mathbf{J}_{\mathbf{F} \circ \mathbf{G}})_{ij} = \frac{\partial(f_i \circ \mathbf{G})}{\partial y_j}$$

For the product of two matrices $\mathbf{C} = \mathbf{AB}$, where $\mathbf{A} \in \mathbb{R}^{n \times m}$ and $\mathbf{B} \in \mathbb{R}^{m \times l}$, the element at indices i, j is computed as:

$$\mathbf{C}_{ij} = \sum_{k=1}^m \mathbf{A}_{ik} \mathbf{B}_{kj}$$

Thus, for the product of the two Jacobian matrices $\mathbf{J}_{\mathbf{F}} \circ \mathbf{G}$ and $\mathbf{J}_{\mathbf{G}}$,

$$\begin{aligned} ((\mathbf{J}_{\mathbf{F}} \circ \mathbf{G})\mathbf{J}_{\mathbf{G}})_{ij} &= \sum_{k=1}^m (\mathbf{J}_{\mathbf{F}} \circ \mathbf{G})_{ik} (\mathbf{J}_{\mathbf{G}})_{kj} \\ &= \sum_{k=1}^m \left. \frac{\partial f_i}{\partial x_k} \right|_{\mathbf{G}} \frac{\partial g_k}{\partial y_j} \end{aligned}$$

As shown in Lemma 3.2, $\mathbf{J}_{\mathbf{F} \circ \mathbf{G}} = (\mathbf{J}_{\mathbf{F}} \circ \mathbf{G})\mathbf{J}_{\mathbf{G}}$, therefore,

$$\begin{aligned} (\mathbf{J}_{\mathbf{F} \circ \mathbf{G}})_{ij} &= [(\mathbf{J}_{\mathbf{F}} \circ \mathbf{G})\mathbf{J}_{\mathbf{G}}]_{ij} \\ \implies \frac{\partial(f_i \circ \mathbf{G})}{\partial y_j} &= \sum_{k=1}^m \left. \frac{\partial f_i}{\partial x_k} \right|_{\mathbf{G}} \frac{\partial g_k}{\partial y_j} \end{aligned} \tag{9}$$

Considering again the functions $X : \mathbb{R}^2 \rightarrow \mathbb{R}$ and $Y : \mathbb{R}^2 \rightarrow \mathbb{R}$ of parameters α, β and $Z : \mathbb{R}^2 \rightarrow \mathbb{R}$ of parameters x and y . Applying the form derived in (9) to the composite function $z : \alpha, \beta \mapsto Z(X(\alpha, \beta), Y(\alpha, \beta))$ leads to

$$\begin{aligned} \frac{\partial z}{\partial \alpha} &= \sum_{k=1}^2 \frac{\partial Z}{\partial x_k} \frac{\partial g_k}{\partial \alpha} = \frac{\partial Z}{\partial x} \frac{\partial X}{\partial \alpha} + \frac{\partial Z}{\partial y} \frac{\partial Y}{\partial \alpha} \\ \frac{\partial z}{\partial \beta} &= \sum_{k=1}^2 \frac{\partial Z}{\partial x_k} \frac{\partial g_k}{\partial \beta} = \frac{\partial Z}{\partial x} \frac{\partial X}{\partial \beta} + \frac{\partial Z}{\partial y} \frac{\partial Y}{\partial \beta} \end{aligned}$$

■

3.5 Laplace's equation

3.5.1 Definition of the Laplacian

Definition 3.2. The *laplacian* operator, denoted Δ , is defined for some scalar field ϕ as $\Delta\phi = \nabla \cdot \nabla\phi$, which when expanded gives:

$$\begin{aligned} \Delta\phi &= \nabla \cdot \nabla\phi \\ &= \nabla \cdot \begin{bmatrix} \partial\phi/\partial x \\ \partial\phi/\partial y \end{bmatrix} \\ &= \frac{\partial^2\phi}{\partial x^2} + \frac{\partial^2\phi}{\partial y^2} \end{aligned}$$

As discussed in Section 3.3, the vector field representing this essay's idealised fluid can be expressed through the gradient of a velocity potential ϕ . The fluid must also satisfy the continuity equation, as established in Section 3.1, thus

$$\begin{aligned}\nabla \cdot \nabla \phi &= 0 \\ \implies \Delta \phi &= 0\end{aligned}$$

This equation is known as Laplace's equation [Lewis et al., 2022]. As will become evident in Section 4.1, it is helpful to define this equation in polar coordinates.

3.5.2 The polar form of the Laplacian

Lemma 3.4 (The polar form of the Laplacian). For the scalar field ϕ , the Laplacian is defined in polar coordinates as

$$\Delta \phi = \frac{\partial^2 \phi}{\partial r^2} + \frac{1}{r} \frac{\partial \phi}{\partial r} + \frac{1}{r^2} \frac{\partial^2 \phi}{\partial \vartheta^2}$$

Proof. The Cartesian coordinates x and y are defined in terms of the polar coordinates r and ϑ as

$$\begin{aligned}x &= r \cos \vartheta \\ y &= r \sin \vartheta\end{aligned}$$

Consequently the partial derivatives for x and y in terms of r and ϑ become

$$\begin{aligned}\frac{\partial x}{\partial r} &= \cos \vartheta & \frac{\partial x}{\partial \vartheta} &= -r \sin \vartheta \\ \frac{\partial y}{\partial r} &= \sin \vartheta & \frac{\partial y}{\partial \vartheta} &= r \cos \vartheta\end{aligned}$$

Therefore, using the chain rule derived in Lemma 3.3, the partial derivative of ϕ in terms of

r becomes:

$$\begin{aligned}\frac{\partial \phi}{\partial r} &= \frac{\partial \phi}{\partial x} \frac{\partial x}{\partial r} + \frac{\partial \phi}{\partial y} \frac{\partial y}{\partial r} \\ &= \frac{\partial \phi}{\partial x} \cos \vartheta + \frac{\partial \phi}{\partial y} \sin \vartheta\end{aligned}\tag{10}$$

And in terms of ϑ (applying the chain rule again),

$$\begin{aligned}\frac{\partial \phi}{\partial \vartheta} &= \frac{\partial \phi}{\partial x} \frac{\partial x}{\partial \vartheta} + \frac{\partial \phi}{\partial y} \frac{\partial y}{\partial \vartheta} \\ &= -\frac{\partial \phi}{\partial x} r \sin \vartheta + \frac{\partial \phi}{\partial y} r \cos \vartheta \\ &= r \left(\frac{\partial \phi}{\partial y} \cos \vartheta - \frac{\partial \phi}{\partial x} \sin \vartheta \right)\end{aligned}\tag{11}$$

Taking the second order partial derivatives with respect to r and using Theorem 2.1 (Clairaut's theorem), which will henceforth be taken for granted, gives

$$\begin{aligned}\frac{\partial^2 \phi}{\partial r^2} &= \frac{\partial}{\partial r} \frac{\partial \phi}{\partial x} \cos \vartheta + \frac{\partial}{\partial r} \frac{\partial \phi}{\partial y} \sin \vartheta \\ &= \frac{\partial}{\partial x} \frac{\partial \phi}{\partial r} \cos \vartheta + \frac{\partial}{\partial y} \frac{\partial \phi}{\partial r} \sin \vartheta\end{aligned}$$

Substituting back from (10),

$$\begin{aligned}\frac{\partial^2 \phi}{\partial r^2} &= \frac{\partial}{\partial x} \left(\frac{\partial \phi}{\partial x} \cos \vartheta + \frac{\partial \phi}{\partial y} \sin \vartheta \right) \cos \vartheta + \frac{\partial}{\partial y} \left(\frac{\partial \phi}{\partial x} \cos \vartheta + \frac{\partial \phi}{\partial y} \sin \vartheta \right) \sin \vartheta \\ &= \frac{\partial^2 \phi}{\partial x^2} \cos^2 \vartheta + \frac{\partial^2 \phi}{\partial y \partial x} (\sin \vartheta) \cos \vartheta + \frac{\partial^2 \phi}{\partial x \partial y} (\cos \vartheta) \sin \vartheta + \frac{\partial^2 \phi}{\partial y^2} \sin^2 \vartheta \\ &= \frac{\partial^2 \phi}{\partial x^2} \cos^2 \vartheta + \frac{\partial^2 \phi}{\partial y^2} \sin^2 \vartheta + \left(\frac{\partial^2 \phi}{\partial y \partial x} + \frac{\partial^2 \phi}{\partial x \partial y} \right) \cos \vartheta \sin \vartheta \\ &= \frac{\partial^2 \phi}{\partial x^2} \cos^2 \vartheta + \frac{\partial^2 \phi}{\partial y^2} \sin^2 \vartheta + 2 \frac{\partial^2 \phi}{\partial y \partial x} \cos \vartheta \sin \vartheta\end{aligned}$$

Applying the same process for the second partial derivative with respect to ϑ leads to

$$\begin{aligned}\frac{\partial^2 \phi}{\partial \vartheta^2} &= \frac{\partial}{\partial \vartheta} \left(\frac{\partial \phi}{\partial y} \cos \vartheta - \frac{\partial \phi}{\partial x} \sin \vartheta \right) r \\ &= \frac{\partial}{\partial \vartheta} \left(\frac{\partial \phi}{\partial y} r \cos \vartheta \right) - \frac{\partial}{\partial \vartheta} \left(\frac{\partial \phi}{\partial x} r \sin \vartheta \right)\end{aligned}$$

$\partial\phi/\partial x$, $\partial\phi/\partial y$, $\cos\vartheta$ and $\sin\vartheta$ are functions of ϑ , so the product rule is applied leading to:

$$\begin{aligned}
\frac{\partial^2\phi}{\partial\vartheta^2} &= \left(\left[\frac{\partial}{\partial\vartheta} \frac{\partial\phi}{\partial y} \right] r \cos\vartheta + \frac{\partial\phi}{\partial y} \left[\frac{\partial}{\partial\vartheta} r \cos\vartheta \right] \right) - \left(\left[\frac{\partial}{\partial\vartheta} \frac{\partial\phi}{\partial x} \right] r \sin\vartheta + \frac{\partial\phi}{\partial x} \left[\frac{\partial}{\partial\vartheta} r \sin\vartheta \right] \right) \\
&= \left(\frac{\partial^2\phi}{\partial y \partial\vartheta} r \cos\vartheta - \frac{\partial\phi}{\partial y} r \sin\vartheta \right) - \left(\frac{\partial^2\phi}{\partial x \partial\vartheta} r \sin\vartheta + \frac{\partial\phi}{\partial x} r \cos\vartheta \right) \\
&= r \left(\frac{\partial}{\partial y} \frac{\partial\phi}{\partial\vartheta} r \cos\vartheta - \frac{\partial}{\partial x} \frac{\partial\phi}{\partial\vartheta} r \sin\vartheta \right) - r \left(\frac{\partial\phi}{\partial x} \cos\vartheta + \frac{\partial\phi}{\partial y} \sin\vartheta \right) \\
&= \underbrace{r^2 \left(\frac{\partial}{\partial y} \frac{\partial\phi}{\partial\vartheta} \cos\vartheta - \frac{\partial}{\partial x} \frac{\partial\phi}{\partial\vartheta} \sin\vartheta \right)}_{\Phi} - r \left(\frac{\partial\phi}{\partial x} \cos\vartheta + \frac{\partial\phi}{\partial y} \sin\vartheta \right) \tag{12}
\end{aligned}$$

Substituting (11) into Φ ,

$$\begin{aligned}
\Phi &= r^2 \left(\frac{\partial}{\partial y} \left(\frac{\partial\phi}{\partial y} \cos\vartheta - \frac{\partial\phi}{\partial x} \sin\vartheta \right) \cos\vartheta - \frac{\partial}{\partial x} \left(\frac{\partial\phi}{\partial y} \cos\vartheta - \frac{\partial\phi}{\partial x} \sin\vartheta \right) \sin\vartheta \right) \\
&= r^2 \left(\left(\frac{\partial^2\phi}{\partial y^2} \cos\vartheta - \frac{\partial^2\phi}{\partial x \partial y} \sin\vartheta \right) \cos\vartheta - \left(\frac{\partial^2\phi}{\partial y \partial x} \cos\vartheta - \frac{\partial^2\phi}{\partial x^2} \sin\vartheta \right) \sin\vartheta \right) \\
&= r^2 \left(\frac{\partial^2\phi}{\partial y^2} \cos^2\vartheta - \frac{\partial^2\phi}{\partial x \partial y} \cos\vartheta \sin\vartheta - \frac{\partial^2\phi}{\partial x \partial y} \cos\vartheta \sin\vartheta + \frac{\partial^2\phi}{\partial x^2} \sin^2\vartheta \right) \\
&= r^2 \left(\frac{\partial^2\phi}{\partial y^2} \cos^2\vartheta + \frac{\partial^2\phi}{\partial x^2} \sin^2\vartheta - 2 \frac{\partial^2\phi}{\partial y \partial x} \cos\vartheta \sin\vartheta \right)
\end{aligned}$$

Substituting Φ back into (12),

$$\frac{\partial^2\phi}{\partial\vartheta^2} = r^2 \left(\frac{\partial^2\phi}{\partial y^2} \cos^2\vartheta + \frac{\partial^2\phi}{\partial x^2} \sin^2\vartheta - 2 \frac{\partial^2\phi}{\partial y \partial x} \cos\vartheta \sin\vartheta \right) - r \left(\frac{\partial\phi}{\partial x} \cos\vartheta + \frac{\partial\phi}{\partial y} \sin\vartheta \right)$$

Substituting in (10) leads to

$$\frac{\partial^2\phi}{\partial\vartheta^2} = r^2 \left(\frac{\partial^2\phi}{\partial y^2} \cos^2\vartheta + \frac{\partial^2\phi}{\partial x^2} \sin^2\vartheta - 2 \frac{\partial^2\phi}{\partial y \partial x} \cos\vartheta \sin\vartheta \right) - r \frac{\partial\phi}{\partial r}$$

Taking the sum of $\partial^2\phi/\partial r^2$ and $(r^{-2}) \partial^2\phi/\partial\vartheta^2$ gives,

$$\begin{aligned}
\frac{\partial^2\phi}{\partial r^2} + \frac{1}{r^2} \frac{\partial^2\phi}{\partial\vartheta^2} &= \left(\frac{\partial^2\phi}{\partial x^2} \cos^2\vartheta + \frac{\partial^2\phi}{\partial y^2} \sin^2\vartheta + 2 \frac{\partial^2\phi}{\partial y \partial x} \cos\vartheta \sin\vartheta \right) \\
&\quad + \left(\frac{\partial^2\phi}{\partial y^2} \cos^2\vartheta + \frac{\partial^2\phi}{\partial x^2} \sin^2\vartheta - 2 \frac{\partial^2\phi}{\partial y \partial x} \cos\vartheta \sin\vartheta - \frac{1}{r} \frac{\partial\phi}{\partial r} \right) \\
&= \frac{\partial^2\phi}{\partial x^2} \cos^2\vartheta + \frac{\partial^2\phi}{\partial x^2} \sin^2\vartheta + \frac{\partial^2\phi}{\partial y^2} \cos^2\vartheta + \frac{\partial^2\phi}{\partial y^2} \sin^2\vartheta - \frac{1}{r} \frac{\partial\phi}{\partial r} \\
&= \frac{\partial^2\phi}{\partial x^2} + \frac{\partial^2\phi}{\partial y^2} - \frac{1}{r} \frac{\partial\phi}{\partial r} \\
\therefore \frac{\partial^2\phi}{\partial r^2} + \frac{1}{r} \frac{\partial\phi}{\partial r} + \frac{1}{r^2} \frac{\partial^2\phi}{\partial\vartheta^2} &= \frac{\partial^2\phi}{\partial x^2} + \frac{\partial^2\phi}{\partial y^2} \\
\therefore \Delta\phi &= \frac{\partial^2\phi}{\partial r^2} + \frac{1}{r} \frac{\partial\phi}{\partial r} + \frac{1}{r^2} \frac{\partial^2\phi}{\partial\vartheta^2}
\end{aligned}$$

■

4 Formulation of the problem

4.1 Boundary equations

5 Conclusion

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8 Appendix

Definition 8.1. The directional derivative in the direction of some vector \mathbf{v} of the function f which is differentiable in the open disk $\mathbb{D}_\delta(\langle x_0, y_0 \rangle)$ centred at (x_0, y_0) with radius $\delta > 0$ is defined as

$$\nabla_{\mathbf{v}} f \Big|_{(x_0, y_0)} = \frac{\nabla f|_{(x_0, y_0)} \cdot \mathbf{v}}{\|\mathbf{v}\|}$$

[Giannakidis and Petrou, 2010]