

Mathematical Methods I

Three dimensional vectors and vector calculus

Define vector $\underline{r} = (q_1, q_2, q_3)$ w.r.t some basis.
Let this basis be related to cartesian (x, y, z) s.t.
we have 3 simultaneous equations for x, y, z in terms of
 q_1, q_2, q_3 . i.e. x, y, z are functions of q_1, q_2, q_3 .

Hence by the chain rule: $dx = \sum_{i=1}^3 \frac{\partial x}{\partial q_i} dq_i$, $dy = \dots$

Now in cartesian: $d\underline{r} = dx \underline{e}_x + dy \underline{e}_y + dz \underline{e}_z$
where $\underline{e}_x, \underline{e}_y, \underline{e}_z$ are basis vectors: unitary and
mutually orthogonal: $\rightarrow$ ORTHONORMAL. So substituting
above result:

$$\begin{aligned} d\underline{r} &= \left(\sum_{i=1}^3 \frac{\partial x}{\partial q_i} dq_i \right) \underline{e}_x + \left(\sum_{i=1}^3 \frac{\partial y}{\partial q_i} dq_i \right) \underline{e}_y + \left(\sum_{i=1}^3 \frac{\partial z}{\partial q_i} dq_i \right) \underline{e}_z \\ &= \sum_{i=1}^3 \left(\frac{\partial x}{\partial q_i} \underline{e}_x + \frac{\partial y}{\partial q_i} \underline{e}_y + \frac{\partial z}{\partial q_i} \underline{e}_z \right) dq_i = \sum_{i=1}^3 \underline{h}_i(q) dq_i \end{aligned}$$

Define $\underline{h}_i(q) = h_i(q) \underline{e}_i(q)$ i.e. $|\underline{e}_i| = 1$

Hence ANY curvilinear coordinate system can be used
to describe a vector in differential form if it can be related
to a cartesian basis. If this coordinate system or BASIS
is orthogonal, i.e. $\underline{e}_i \cdot \underline{e}_j = \delta_{ij}$ then:

(for basis (q_1, q_2, q_3)):

$$d\underline{r} = \sum_{i=1}^3 h_i dq_i \underline{e}_i$$

where

$$h_i^2 = \left(\frac{\partial x}{\partial q_i} \right)^2 + \left(\frac{\partial y}{\partial q_i} \right)^2 + \left(\frac{\partial z}{\partial q_i} \right)^2$$

$$(i.e. \underline{h}_i \cdot \underline{h}_i = h_i^2 (\underline{e}_i \cdot \underline{e}_i) = h_i^2)$$

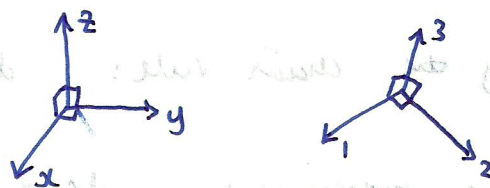
Using this new basis we can define a volume element

$d\tau$ as :

Vector triple product
= volume of a parallelepiped.

$$\begin{aligned} d\tau &= (dr_1 \times dr_2) \cdot dr_3 \\ &= (h_1 e_1 dr_1 \times h_2 e_2 dr_2) \cdot h_3 e_3 dr_3 \\ &= h_1 h_2 h_3 dr_1 dr_2 dr_3 (e_1 \times e_2) \cdot e_3 \end{aligned}$$

Now $e_1 \times e_2 = e_3$ in R.H. basis of curvilinear



$$d\tau = h_1 h_2 h_3 dr_1 dr_2 dr_3$$

Now for a scalar field ϕ of x, y, z [curvilinear]

$$d\phi = \frac{\partial \phi}{\partial x} dx + \frac{\partial \phi}{\partial y} dy + \frac{\partial \phi}{\partial z} dz = \nabla \phi \cdot dr$$

i.e. vector operator ∇ is defined in curvilinear by :

$$\nabla = \underline{i} \frac{\partial}{\partial x} + \underline{j} \frac{\partial}{\partial y} + \underline{k} \frac{\partial}{\partial z}$$

Interpretation: For surfaces of constant ϕ $d\phi = 0 \Rightarrow \nabla \phi$ or $dr = 0$. (or both). Since dr could represent any displacement in ϕ ; $d\phi = 0 \Rightarrow$ two characteristics of $\nabla \phi$.

(1) $\nabla \phi \perp dr$ i.e. $\nabla \phi$ points normal to any surface of constant ϕ .

$\Rightarrow$ (2) If surface is a local stationary region then since dr could be anything; $d\phi = 0 \Rightarrow \nabla \phi = 0$. So $\nabla \phi = 0$ is the equation of a stationary point. Now if $d\phi \neq 0$

clearly it is maximised when $\nabla \phi \parallel dr$. i.e. $\nabla \phi$ points in the direction where ϕ is changing most rapidly. Hence the name Gradient.

Curl and divergence of vector fields generated by $\underline{F} = \underline{F}(\underline{r})$ are given by:

$$\text{curl } \underline{F} = \underline{\nabla} \times \underline{F} \quad \text{div } \underline{F} = \underline{\nabla} \cdot \underline{F}$$

← Vector
← Scalar

NOTE curl grad = 0 i.e., $\underline{\nabla} \times \underline{\nabla} = 0$

Now $\underline{\nabla} \cdot \underline{\nabla} \phi \equiv \nabla^2 \phi$ which is the (scalar) Laplacian.

We can write $\underline{\nabla}$, curl, div, ∇^2 in general curvilinear coordinates using $d\phi = \underline{\nabla} \phi \cdot d\underline{r}$ as the defining property of $\underline{\nabla}$. For $\phi(r_1, r_2, r_3)$:

$$\text{i.e., } d\phi = \sum_{i=1}^3 \frac{\partial \phi}{\partial r_i} dr_i = \sum_{i=1}^3 \left(\frac{1}{h_i} \frac{\partial \phi}{\partial r_i} \right) h_i dr_i$$

$$\text{Now since } d\underline{r} = \sum_{i=1}^3 h_i dr_i \underline{e}_i$$

$$\Rightarrow \text{as } d\phi = \underline{\nabla} \phi \cdot d\underline{r} \quad ; \quad \underline{\nabla} \phi = \sum_{i=1}^3 \left(\frac{1}{h_i} \frac{\partial \phi}{\partial r_i} \right) \underline{e}_i$$

$$\text{i.e. } \underline{\nabla} = \sum_{i=1}^3 \frac{\underline{e}_i}{h_i} \frac{\partial}{\partial r_i}$$

using this result plus $\underline{\nabla} \times \underline{\nabla} = 0$ and other vector formulae: We can generate

the following table of vector operators in curvilinear coordinates.

... scalar	grad	$\underline{\nabla} = \sum_{i=1}^3 \frac{\underline{e}_i}{h_i} \frac{\partial}{\partial r_i} \dots$	(operates on scalars)
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... vector	div	$\underline{\nabla} \cdot \dots = \frac{1}{h_1 h_2 h_3} \left[\frac{\partial}{\partial r_1} (h_2 h_3 \dots)_1 + \frac{\partial}{\partial r_2} (h_3 h_1 \dots)_2 + \frac{\partial}{\partial r_3} (h_1 h_2 \dots)_3 \right]$
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... vector	curl	$\underline{\nabla} \times \dots = \frac{1}{h_1 h_2 h_3} \begin{vmatrix} h_1 \underline{e}_1 & h_2 \underline{e}_2 & h_3 \underline{e}_3 \\ \frac{\partial}{\partial r_1} & \frac{\partial}{\partial r_2} & \frac{\partial}{\partial r_3} \\ h_1 \dots_1 & h_2 \dots_2 & h_3 \dots_3 \end{vmatrix}$
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... scalar 2	Laplacian	$\nabla^2 \dots = \frac{1}{h_1 h_2 h_3} \left[\frac{\partial}{\partial r_1} \left(\frac{h_2 h_3}{h_1} \frac{\partial}{\partial r_1} \dots \right) + \frac{\partial}{\partial r_2} \left(\frac{h_1 h_3}{h_2} \frac{\partial}{\partial r_2} \dots \right) + \frac{\partial}{\partial r_3} \left(\frac{h_1 h_2}{h_3} \frac{\partial}{\partial r_3} \dots \right) \right]$
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Vector operators are made especially useful by their presence in Stokes' and Divergence theorems.

Stokes' Theorem (Generalised).

For open surface S bounded by closed curve C :

$$\oint_C d\mathbf{l} \cdot \left\{ \begin{array}{c} \cdot \underline{A} \\ \times \underline{A} \\ \phi \end{array} \right\} = \int_S (d\mathbf{s} \times \nabla) \cdot \left\{ \begin{array}{c} \cdot \underline{A} \\ \times \underline{A} \\ \phi \end{array} \right\}$$

$d\mathbf{l}$ is a line element of C , $d\mathbf{s}$ is a surface element of S , $\underline{A}$ is a vector field and ϕ is a scalar field assumed present in the vicinity of S .

$$\text{i.e., } \oint_C d\mathbf{l} \cdot \underline{A} = \int_S (d\mathbf{s} \times \nabla) \cdot \underline{A} = \int_S (\nabla \times \underline{A}) \cdot d\mathbf{s}$$

Divergence Theorem (Generalised)

For a closed surface S having volume V :

$$\oint_S d\mathbf{s} \cdot \left\{ \begin{array}{c} \cdot \underline{A} \\ \times \underline{A} \\ \phi \end{array} \right\} = \int_V d\tau \nabla \cdot \left\{ \begin{array}{c} \cdot \underline{A} \\ \times \underline{A} \\ \phi \end{array} \right\}$$

$d\tau$ is a volume element of V .

$$\text{i.e., } \oint_S d\mathbf{s} \cdot \underline{A} = \int_V d\tau \nabla \cdot \underline{A}$$

Integral Transforms : Fourier (f) and Laplace (L)

Fourier transforms

A function, $f(x)$ may be expressed as a Fourier series of harmonic functions of period T .

$$f(x) = \sum_{n=-\infty}^{\infty} c_n e^{2\pi i n x / T} = \sum_{n=-\infty}^{\infty} c_n e^{i \omega_n x}$$

where $\omega_n = \frac{2\pi n}{T}$ so as $T \rightarrow \infty$; if $f(x)$ fulfils the DIRICHLET conditions:

- (1) $f(x)$ periodic with period $T \leftarrow$ Always ok if $T \rightarrow \infty$!
- (2) Single valued and continuous
- except at a FINITE number of finite discontinuities.
- (3) Finite number of stationary points (max, min) within one period
- (4) $\int_{-T/2}^{T/2} |f(x)| dx$ is convergent, i.e., $\neq \infty$.

then changes in ω_n become vanishingly small
 $\rightarrow \omega_n$ becomes continuous. ($\Delta\omega = 2\pi/T \rightarrow 0$).

Now as $c_n = \frac{1}{T} \int_{-T/2}^{T/2} f(x) e^{-2\pi i n x / T} dx = \frac{\Delta\omega}{2\pi} \int_{-T/2}^{T/2} f(x) e^{-i \omega x} dx$

$$\Rightarrow f(x) = \sum_{n=-\infty}^{\infty} \frac{\Delta\omega}{2\pi} \int_{-T/2}^{T/2} f(u) e^{-i \omega u} du e^{i \omega x}$$

dummy variable u - no x dependence of c_n
(only ω is varied. don't want to confuse with x !).

Now as $T \rightarrow \infty$, $\Delta\omega \rightarrow 0$ so first sum $\rightarrow$ integral. ($\Delta\omega \rightarrow d\omega$).

$$\text{i.e., } f(x) = \frac{1}{2\pi} \int_{-\infty}^{\infty} d\omega e^{i \omega x} \int_{-\infty}^{\infty} f(u) e^{-i \omega u} du \quad (*)$$

Now if we define the Fourier Transform of $f(x)$, $\tilde{f}(\omega)$ as:

$$\tilde{f}(\omega) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} f(x) e^{-i\omega x} dx$$

then (*) $\Rightarrow f(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \tilde{f}(\omega) e^{i\omega x} d\omega$

This is Fourier's inversion theorem.

Proof: $f(x) = \frac{1}{2\pi} \int_{-\infty}^{\infty} d\omega e^{i\omega x} \int_{-\infty}^{\infty} du f(u) e^{-i\omega u}$ TRUE. $\neq$

Now defining $\tilde{f}(\omega)$ above: let $f(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \tilde{f}(\omega) e^{i\omega x} d\omega$

$$\Rightarrow f(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \left[\frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} f(x) e^{-i\omega x} dx \right] e^{i\omega x} d\omega$$

$\Rightarrow$ use dummy variables to not confuse integrals over $e^{\pm i\omega x}$

$$\Rightarrow f(x) = \frac{1}{2\pi} \int_{-\infty}^{\infty} d\omega e^{i\omega x} \int_{-\infty}^{\infty} f(u) e^{-i\omega u} du \quad \text{i.e. (*)}$$

So the Fourier inversion theorem works.

Important functions that behave in interesting ways when Fourier Transformed.

(1) Gaussian. $f(x) = \frac{1}{\sigma\sqrt{2\pi}} \exp(-x^2/2\sigma^2)$

$$f[f(x)] = \tilde{f}(\omega) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{\sigma^2 \omega^2}{2}\right) \quad \text{i.e. another Gaussian.}$$

Note 'width' of Gaussian = σ and $1/\sigma$ for FT.

$\Rightarrow \Delta\omega \Delta x = 1$ where ' Δ ' $\Rightarrow$ 'spread' in ω and x .

(2) Dirac δ function

The Dirac δ function is defined:

$$\int_{-\infty}^{\infty} f(x) \delta(x-a) dx = f(a)$$

$f(x) \delta(x-a) = 0, x \neq a$
So at $x=a$, $f(x) = \text{value}$
So is not integrated.

i.e., $\delta(x) = 0, x \neq 0$ and $\int_{-\infty}^{\infty} \delta(x) dx = 1$

(An infinitely sharp pulse at some 'origin' with unit integral over all space).

Now it is clear from the definition that $\delta(x)$ is REAL and SYMMETRIC. i.e., $\delta(x) = \delta(-x)$ and $\delta^*(x) = \delta(x)$.

Now $f[\delta(x)] = \tilde{f}(\omega) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-i\omega x} \delta(x) dx$

Now $e^{-i\omega x} \delta(x) = 0$ when $x \neq 0$; $e^0 = 1$ when $x=0$

So $\tilde{f}(\omega) = \frac{1}{\sqrt{2\pi}}$

$\delta(x)$ is related to two other functions.

(1) Heaviside step function $H(x)$.

$$H(x) = \begin{cases} 0 & x < 0 \\ 1 & x \geq 0 \end{cases}$$



$H'(x) = \delta(x)$

Proof: $f(x) = \int_{-\infty}^{\infty} \delta(x) f(x) dx = \int_{-\infty}^{\infty} H'(x) f(x) dx$ is true.

i.e., $H'(x)$ has the same property of 'picking out $f(x)$ ' as $\delta(x)$ hence the equality.

(2) Sinc function, $f(x) = \frac{\sin x}{x}$

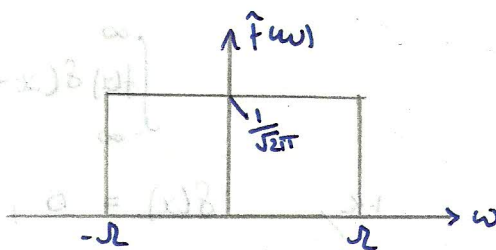
Consider a Fourier transform $\tilde{f}(\omega)$ which is a top hat of width 2Ω and height $\frac{1}{\sqrt{2\pi}}$

$$\text{Now } f(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \tilde{f}(\omega) e^{i\omega x} d\omega$$

$$= \frac{1}{\sqrt{2\pi}} \int_{-\Omega}^{\Omega} \frac{1}{\sqrt{2\pi}} e^{i\omega x} d\omega$$

$$= \frac{1}{2\pi} \left[\frac{1}{ix} e^{i\omega x} \right]_{-\Omega}^{\Omega}$$

$$= \frac{1}{2\pi} \left[\frac{1}{ix} e^{i\Omega x} - \frac{1}{ix} e^{-i\Omega x} \right] = \frac{\sin \Omega x}{\pi}$$



Now if $\Omega \rightarrow \infty$, $\tilde{f}(\omega) \rightarrow \tilde{f}(\omega) = \frac{1}{\sqrt{2\pi}} \quad |\omega| \leq \infty$.

which is the 'definition' of the $\delta(x)$. Hence

$$\delta(x) = \lim_{\Omega \rightarrow \infty} \left(\frac{\sin \Omega x}{\pi} \right)$$

Properties and applications of Fourier Transforms.

(1) Differentiation

$$f[f^{(n)}(x)] = (i\omega)^n \tilde{f}(\omega)$$

(2) Integration

$$f\left[\int f(x) dx\right] = \frac{1}{i\omega} \tilde{f}(\omega) + 2\pi c \delta(\omega)$$

(= integration constant)

(3) Scaling

$$f[ta(x)] = \frac{1}{a} \tilde{f}\left(\frac{\omega}{a}\right)$$

(4) Translation

$$f[f(x+a)] = e^{i\omega a} \tilde{f}(\omega)$$

(5) Exponential multiplication

$$f[e^{\alpha x} f(x)] = \tilde{f}(\omega + i\alpha)$$

Convolution

For functions $f(x)$ and $g(x)$ (' f ' and ' g ') :
convolution $f * g$ is defined by:

$$f * g = \int_{-\infty}^{\infty} dy f(y) g(x-y)$$

(y again
a dummy variable
to change in
form x).

Note $f * g = g * f$.

Convolution theorem:

$$f[f * g] = \frac{1}{\sqrt{2\pi}} \tilde{f} * \tilde{g}$$

AND

$$f[f * g] = \sqrt{2\pi} \tilde{f} \tilde{g}$$

This is handy in the theory of optics/waves.

Parseval's Theorem

$$\int_{-\infty}^{\infty} |f(x)|^2 dx = \int_{-\infty}^{\infty} |\tilde{f}(\omega)|^2 d\omega$$

useful for evaluating definite integrals over all space
where FT is integrable and original function is not
/ need to integrate.

Laplace Transform

For functions defined $x > 0$
and have convergence problems
when $x \rightarrow \infty$ - another integral transform is used instead
of Fourier transforms. (Non convergent as $x \rightarrow \infty$ infringes
the Dirichlet conditions). This is the Laplace Transform.

$$\tilde{f}(s) = \int_0^{\infty} f(x) e^{-sx} dx$$

Inverse is NOT easy to find. (No 'Laplace Inversion
theorem'). $\rightarrow$ Require star integration in complex plane.
 $\rightarrow$ Use comparison with standard table of LT's for solutions.

Laplace Transforms are useful in solving ODE's by reducing them to linear equations. These can be compared to a table of LT's for the solution to be read off.

This makes use of the property:

$$\mathcal{L}[f^{(n)}(x)] = s^n \bar{f}(s) - s^{n-1} f(0) - s^{n-2} f'(0) - \dots - f^{(n-1)}(0)$$

other properties and 'standard' Laplace Transforms.

Function $f(x)$ $x > 0$	LT. $\bar{f}(s)$
$\delta(x)$	1
$H(x)$	$1/s$
x^n	$n! / s^{n+1}$
$\sqrt{x}$	$\frac{1}{2} (\pi / s^3)^{1/2}$
$x^{1/2}$	$(\pi / s)^{1/2}$
e^{-ax}	$1 / (s+a)$
$\sin \omega x$	$\omega / (s^2 + \omega^2)$
$\cos \omega x$	$s / (s^2 + \omega^2)$
$\sinh \omega x$	$\omega / (s^2 - \omega^2)$
$\cosh \omega x$	$s / (s^2 - \omega^2)$
$e^{-ax} f(x)$	$\bar{f}(s+a)$
$f(x-\tau) H(x-\tau)$	$e^{-s\tau} \bar{f}(s)$
$x f(x)$	$-d\bar{f}(s)/ds$
$\int_0^x f(\tau) d\tau$	$\bar{f}(s)/s$

Convolution theorem for Laplace Transforms: For f, g

$$\mathcal{L}[f * g] = \mathcal{L}[g * f] = \bar{f} \bar{g}$$

Note inverse is not true because the

when g is a Laplace Transform inversion theorem.

Matrices and Vector spaces

In general terms a vector is a single columned matrix with N rows corresponding to the N DIMENSIONS of the VECTOR SPACE that we use to describe it.

A vector space of N dimensions is defined by a set of N LINEARLY INDEPENDANT vectors which form a BASIS for the vector space.

linearly independent vector sets $\{\underline{u}_1, \underline{u}_2, \dots, \underline{u}_N\}$ that form an N dimensional basis satisfy the following:

$$\text{If } \sum_{i=1}^N a_i \underline{u}_i = \underline{0} \Rightarrow a_i = 0 \text{ for scalars } a_i.$$

Hence if we introduce another vector $\underline{v}$ into the vector space it becomes linearly dependant and $\therefore$ we can write $\underline{v}$ in terms of all the other vectors

$$\text{i.e., } \underline{v} = \sum_{i=1}^N a_i \underline{u}_i$$

linear independence of $\{\underline{u}_i\} \Rightarrow \{a_i\}$ must be unique so $\underline{v}$ is uniquely defined by $\sum_{i=1}^N a_i \underline{u}_i$.

so we can express any vector $\underline{v}$ in terms of its components and the basis of the vector space that $\underline{v}$ exists in.

i.e. for 3D Euclidean (Cartesian) vector space

$$\underline{v} = v_x \underline{e}_x + v_y \underline{e}_y + v_z \underline{e}_z$$

basis = $\{\underline{e}_x, \underline{e}_y, \underline{e}_z\}$.

Note a vector is independent of the basis used to describe it. The reality of a vector (in 3D a line of finite magnitude and some concept of direction - an arrow if you like) is independent of description though its existence must be defined by some basis which can be arbitrary. Sounds a bit philosophical....

Scalar product is an operation between two vectors which yields a scalar.

denoted as $\underline{u} \cdot \underline{v}$ for vectors $\underline{u}, \underline{v}$.

- Properties:
- (1) $\underline{u} \cdot \underline{v} \in \mathbb{C}$
 - (2) $(\underline{u} \cdot \underline{v})^* = (\underline{v} \cdot \underline{u})$
 - (3) $\underline{u} \cdot (a\underline{v}_1 + b\underline{v}_2) = a\underline{u} \cdot \underline{v}_1 + b\underline{u} \cdot \underline{v}_2$
 - (4) $\underline{v} \cdot \underline{v} \geq 0$ i.e. can write $\underline{v} \cdot \underline{v} = |\underline{v}|^2$
 - (5) $|\underline{v}| = 0 \Rightarrow \underline{v} = 0$

Note $(a\underline{u}_1 + b\underline{u}_2) \cdot \underline{v} = a^* \underline{u}_1 \cdot \underline{v} + b^* \underline{u}_2 \cdot \underline{v}$
 and $a\underline{u} \cdot b\underline{v} = a^* b \underline{u} \cdot \underline{v}$

So scalar product of two vectors $\underline{u} = \sum_{i=1}^N a_i \underline{e}_i$
 $\underline{b} = \sum_{j=1}^N b_j \underline{e}_j$ w.r.t N dimensional vector space:
 $\{\underline{e}_j\} \leftarrow$ basis vector set.

$$\underline{u} \cdot \underline{b} = \left(\sum_{i=1}^N a_i \underline{e}_i \right) \cdot \left(\sum_{j=1}^N b_j \underline{e}_j \right)$$

$$= \sum_{i=1}^N \sum_{j=1}^N a_i^* b_j \underline{e}_i \cdot \underline{e}_j$$

Now if basis is orthogonal: $\underline{e}_i \cdot \underline{e}_j = \delta_{ij}$

Hence $\underline{u} \cdot \underline{b} = \sum_{i=1}^N a_i^* b_i$

Now if $\underline{e}_i \cdot \underline{e}_j \neq \delta_{ij}$ then define a matrix $\underline{G}$ such that $G_{ij} = \underline{e}_i \cdot \underline{e}_j$.

$$\text{So } \underline{a} \cdot \underline{b} = \sum_{i=1}^N \sum_{j=1}^N a_i^* G_{ij} b_j$$

Now although we have introduced the 'dummy suffix' j to avoid confusion with i the above $\Rightarrow$ we can interpret $\underline{a} \cdot \underline{b}$ as a matrix multiplication with i, j corresponding to rows, columns.

$$\underline{a} \cdot \underline{b} = \underline{a}^+ \underline{G} \underline{b} \quad \underline{a}^+ = (a_1^* \ a_2^* \ a_3^* \ \dots \ a_n^*)$$

So for orthogonal basis: $\underline{G} = \underline{I} \Rightarrow \underline{a} \cdot \underline{b} = \underline{a}^+ \underline{b}$

To ensure correct properties of scalar products: $\underline{G}$ must be Hermitian. i.e. $\underline{G} = \underline{G}^+$

Note if vector space is real then all vectors described by it are real. Hence $\underline{G}$ is also real.

$\therefore$ as $\underline{G}^+ = \underline{G}$ and $\underline{G}^* = \underline{G} \Rightarrow \underline{G} = \underline{G}^T$ which is the defining property of a real symmetric matrix.

Matrices and their properties

Matrices are Associative but non-commutative (usually).

They multiply according to the rule: $\underline{A} \underline{B} = \underline{C}$

$$C_{ij} = (\underline{A} \underline{B})_{ij} = \sum_{k=1}^L A_{ik} B_{kj} \quad \text{for } \underline{A} = n \times L \quad \underline{B} = L \times m$$

and $\underline{C}$ is $n \times m$ order.

One can perform operation on matrix elements: for matrix $\underline{\underline{A}}$ ($n \times m$):

$$\underline{\underline{A}} \rightarrow \underline{\underline{A}}^* \Rightarrow A_{ij} \rightarrow A_{ij}^*$$

$$\underline{\underline{A}} \rightarrow \underline{\underline{A}}^T \Rightarrow A_{ij} \rightarrow A_{ji}$$

$$\underline{\underline{A}} \rightarrow \underline{\underline{A}}^\dagger \Rightarrow A_{ij} \rightarrow A_{ji}^*$$

and for square

$$\text{matrices: } \underline{\underline{A}} \rightarrow \underline{\underline{A}}^\dagger$$

$$\Rightarrow A_{ij} = C_{ji}^T \quad (\text{Transpose of cofactors})$$

* see below.

For square ($n \times n$) matrices we have further properties and special types of matrices. let $\underline{\underline{A}} = n \times n$ square matrix.

$$\underline{\underline{A}} \underline{\underline{A}}^{-1} = \underline{\underline{A}}^{-1} \underline{\underline{A}} = \underline{\underline{I}} \quad ; \quad I_{ij} = \delta_{ij}$$

$$\underline{\underline{A}}^{-1} \text{ is called the inverse. Now } A_{ij}^{-1} = \frac{C_{ji}^T}{|\underline{\underline{A}}|}$$

C_{ij} is the 'determinant' of a matrix formed from the reduced/minor matrix formed when row i and column j are removed from the matrix.

$$\text{'determinant'} = \text{minor determinant} \times (-1)^{i+j}$$

Determinant of a 2×2 matrix is defined as:

$$\text{For } \underline{\underline{A}} = \begin{pmatrix} a & b \\ c & d \end{pmatrix} : |\underline{\underline{A}}| = ad - bc.$$

For higher orders of matrices we use a Laplace expansion to find the determinant in terms of 2×2 ones.

$$|\underline{\underline{A}}| = \sum_{j=1}^n a_{ij} C_{ij} \quad (\text{This is one way of doing it}).$$

Each C_{ij} can be found by considering a Laplace expansion of the matrix $\underline{\underline{C}}^i$ (ignore $(-1)^{i+j}$ coeff.) $|\underline{\underline{A}}|$ is $\therefore$ evaluated as a sum of 2×2 determinants.

Inverses of matrices $\underline{A} \underline{B} \underline{C} \dots \underline{Z}$ have the property

$$(\underline{A} \underline{B} \underline{C} \dots \underline{Z})^{-1} = \underline{Z}^{-1} \dots \underline{C}^{-1} \underline{B}^{-1} \underline{A}^{-1}$$

Determinants of matrices $\underline{A} \underline{B} \underline{C} \dots \underline{Z}$ have the property

$$|\underline{A} \underline{B} \underline{C} \dots \underline{Z}| = |\underline{A}| |\underline{B}| |\underline{C}| \dots |\underline{Z}|$$

Special matrices:

(1) Symmetric

$$\underline{A} = \underline{A}^T$$

(2) Anti-symmetric

$$\underline{A} = -\underline{A}^T$$

we can express any matrix $\underline{M}$ as a sum of symmetric and anti-symmetric matrices: ↑ square

$$\underline{M} = \underbrace{\frac{1}{2} (\underline{M} + \underline{M}^T)}_{\text{Symmetric}} + \underbrace{\frac{1}{2} (\underline{M} - \underline{M}^T)}_{\text{Antisymmetric}}$$

(3) Real

$$\underline{A} = \underline{A}^*$$

(4) Pure imaginary

$$\underline{A} = -\underline{A}^*$$

(5) Orthogonal

$$\underline{A}^{-1} = \underline{A}^T$$

($\underline{A}^T$ is also orthogonal and $|\underline{A}| = \pm 1$).

(6) Hermitian

$$\underline{A} = \underline{A}^\dagger$$

anti - "

$$\underline{A} = -\underline{A}^\dagger$$

$$\text{As: } \underline{M} = \frac{1}{2} (\underline{M} + \underline{M}^\dagger) + \frac{1}{2} (\underline{M} - \underline{M}^\dagger)$$

↑
Hermitian

↑
anti-Hermitian

we can write any matrix $\underline{M}$ (square) as a sum of Hermitian and anti-Hermitian matrices.

(3) Unitary $A^\dagger = A^{-1}$
 (8) Normal $A A^\dagger = A^\dagger A$

Eigenvectors and Eigenvalues of a square matrix

For a square matrix $\underline{M}$: $\underline{M} \underline{e} = \lambda \underline{e}$
 defines scalar eigenvalues λ and eigenvectors $\underline{e}$.

Now since $\lambda \underline{e} = \lambda \underline{I} \underline{e} \Rightarrow \underline{M} \underline{e} - \lambda \underline{I} \underline{e} = 0$

Hence $(\underline{M} - \underline{I} \lambda) \underline{e} = 0$ and since $\underline{e}$ is non zero vector

$\Rightarrow |\underline{M} - \underline{I} \lambda| = 0$ This characteristic equation yields a polynomial of order of $\underline{M}$ which allows $\{\lambda\}$ to be found. (There will be a maximum of N λ 's if N is the order of $\underline{M}$).

Now let $\underline{A}$ = matrix of eigenvectors of a matrix $\underline{M}$.
 $(\underline{A})_{ij} = e_{ij}$ i.e. j th eigenvector, component i .

$$\underline{A} = \begin{pmatrix} e_{(1)1} & e_{(2)1} & \dots & e_{(n)1} \\ e_{(1)2} & e_{(2)2} & \dots & e_{(n)2} \\ e_{(1)3} & e_{(2)3} & \dots & e_{(n)3} \\ \vdots & \vdots & \ddots & \vdots \\ e_{(1)n} & e_{(2)n} & \dots & e_{(n)n} \end{pmatrix}$$

Now as $\underline{M} \underline{e} = \lambda \underline{e} \Rightarrow \sum_{k=1}^n M_{jk} e_{ik} = \lambda_i e_{ij}$

$\Rightarrow \sum_{k=1}^n M_{jk} A_{ki} = A_{ji} \lambda_i = \sum_{k=1}^n A_{jk} \delta_{ki} \lambda_i$

$\Rightarrow \underline{M} \underline{A} = \underline{A} \underline{\Lambda}$ where $(\underline{\Lambda})_{ij} = \lambda_i \delta_{ij}$
 Better: $(\underline{M} \underline{A})_{ij} = \sum_k M_{ik} A_{kj} = \sum_k M_{ik} e_{kj} = \sum_k M_{ik} \delta_{ki} \lambda_i = \lambda_i A_{ij} = \sum_k A_{ik} \delta_{ki} \lambda_i$
 Now $\underline{M} \underline{e}_{(i)} = \lambda_i \underline{e}_{(i)} \Rightarrow \sum_k M_{ik} e_{kj} = \lambda_i e_{ij} = \lambda_i A_{ij}$

Let $A_{ij} = e_{ij}$

So any square matrix $\underline{M}$ can be written in the form

$$\underline{M} = \underline{A} \underline{D} \underline{A}^{-1} \quad (*)$$

or equivalently, any square matrix $\underline{M}$ can be diagonalised by $\underline{A}^{-1} \underline{M} \underline{A} = (\underline{D})$.

Now (*) is useful as $\underline{M}^n = \underline{A} \underline{D}^n \underline{A}^{-1}$.

$\underline{D}^n$ is easily evaluated because it is diagonal.

$$(\underline{D}^n)_{ij} = \lambda_{ij}^n \delta_{ij}$$

Hermian matrices have interesting / specific eigenvector, eigenvalue properties. ($\underline{H} = \text{Hermian matrix}$, $\underline{H} = \underline{H}^\dagger$).

- (1) eigenvalues are real
- (2) eigenvectors are orthogonal if eigenvalues are different. ('orthogonal' $\Rightarrow \underline{e}_{ij} \cdot \underline{e}_{il} = 0$).
- (3) $\underline{e}_{ij}$ and $\underline{e}_{il}$ are linearly independent.
- (4) If $\{\underline{e}_{ij}\}$ are orthonormal eigenvectors of $\underline{H}$ then $\underline{A} = [\underline{e}_{(1)} \underline{e}_{(2)} \underline{e}_{(3)} \dots]$ is unitary.

Eigenvectors and Principle axis

For equation $\underline{r}^T \underline{A} \underline{r} = c$.

$c = \text{constant}$; $\underline{r} = \begin{pmatrix} x \\ y \end{pmatrix}$; $\underline{A} = \begin{pmatrix} \alpha & \beta \\ \beta & \gamma \end{pmatrix}$ i.e. real symmetric
 $\Rightarrow \alpha x^2 + 2\beta xy + \gamma y^2 = c$ (equation of conic section).

change basis: $\underline{r} = \underline{R} \underline{r}'$ where $\underline{r}' = \begin{pmatrix} x' \\ y' \end{pmatrix}$ $\rightarrow$ more

and $\underline{R}$ is a real orthogonal rotation matrix

$\Rightarrow$ by substitution: $(\underline{R} \underline{r}')^T \underline{A} \underline{R} \underline{r}' = \underline{r}'^T \underline{R}^T \underline{A} \underline{R} \underline{r}'$

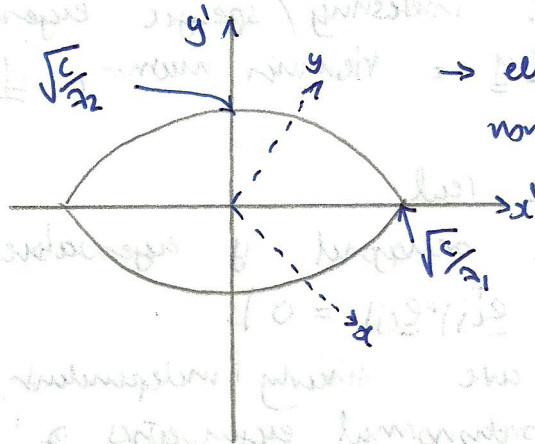
Now since $\underline{R}$ is orthogonal: $\underline{R}^T = \underline{R}^{-1}$

so if $(\underline{R} \underline{r}') = (\{e_{ij}\})$ where $\{e_{ij}\}$ are the eigenvectors of $\underline{A}$ then $\underline{R}^T \underline{A} \underline{R} = \underline{D}$

i.e., $D_{ij} = \lambda_{ij} \delta_{ij}$ where $\{\lambda_{ij}\} = \text{eigenvalues of } \underline{A}$.

So in new basis our becomes $\underline{r}'^T \underline{D} \underline{r}' = C$

i.e., $\underline{\lambda_1 x'^2 + \lambda_2 y'^2 = C}$ which is much easier to visualise.



$\rightarrow$ ellipse if λ_1, λ_2 are +ve and non zero.

Note $\underline{r}$ could be in higher dimensions where in that case we have 3 eigenvalues and a singular.

Because of the squared dependent on the new basis these surfaces are called 'Quadratic Surfaces'.

Now eigenvectors of $\underline{A}$ are aligned with its 'principal axis' - i.e. points where the distance to the quadratic surface from the origin is 'stationary'. (i.e. min/max).

This leads to the expression

$\lambda(e) = \frac{e^T \underline{A} e}{e^T e}$ which computes the eigenvalues

of symmetric matrix A gives the eigenvectors. If $\{e_j\}$ are known this may be a quicker method of finding $\{\lambda_j\}$ than solving the characteristic equation.

For Hermitian matrices, $\frac{1}{2}$

$$\lambda(e) = \frac{e^+ H e}{e^+ e}$$

Infinite series and convergence

Given an infinite sequence of complex numbers $u_1, u_2, \dots$ define partial sum S_N by

$$S_N = \sum_{n=1}^N u_n$$

If $\lim_{N \rightarrow \infty} S_N$ exists and is finite then

$$S = \sum_{n=1}^{\infty} u_n \text{ converges.}$$

ABSOLUTE convergence is if $\sum_{n=1}^{\infty} |u_n|$ converges also otherwise convergence is conditional on the sign of u_n .

i.e. $S = 1 - \frac{1}{2} + \frac{1}{3} - \frac{1}{4} \dots$ converges to $\ln 2$.
but $1 + \frac{1}{2} + \frac{1}{3} + \frac{1}{4} \dots$ is not convergent.

Tests for convergence

(1) Comparison test: If $\sum_{n=1}^{\infty} v_n$ converges then series

$S = \sum_{n=1}^{\infty} u_n$ converges if $|u_n| < C|v_n|$ for some n independent C .

Further tests exploit proven convergence of a geometric series:

$$S = 1 + x + x^2 + x^3 + \dots \quad |x| < 1 \Rightarrow S \neq \infty.$$

Now pr $S_N = 1 + x + \dots + x^{N-1} = \frac{1-x^N}{1-x}$

Proof [Induction] $S_0 = \frac{1-x^0}{1-x} = 0$ $S_1 = \frac{1-x}{1-x} = 1 \checkmark$

↑
Not essential
but helpful!

Assume k th term is true, i.e. $S_k = \frac{1-x^k}{1-x}$

Now S_{k+1} is assumed to be $S_{k+1} = \frac{1-x^{k+1}}{1-x}$

Now we know $S_{k+1} - S_k = x^k$ as $k+1$ th term in series is $x^{k+1-1} = x^k$.

Hence $S_{k+1} - S_k = \frac{1-x^{k+1} - 1 + x^k}{1-x} = \frac{x^k(-x+1)}{1-x}$

$= x^k \checkmark$ so $S_N = \frac{1-x^N}{1-x}$ is true.

Now $\lim_{N \rightarrow \infty} S_N = \frac{1}{1-x}$ pr $|x| < 1$. So Geometric series is convergent pr $|x| < 1$.

(2) Ratio test S converges if $\lim_{n \rightarrow \infty} \left| \frac{u_{n+1}}{u_n} \right| < 1$

(3) Cauchy's test S " " $\lim_{n \rightarrow \infty} |u_n|^{1/n} < 1$

Note these tests test for absolute convergence.

Taylor series. (For $f(z) = \sum_{n=0}^{\infty} a_n z^n$).

If $\sum_{n=0}^{\infty} a_n z^n$ converges for $z = c$ [$c \in \mathbb{C}$] then it absolutely converges for all z s.t. $|z| < |c|$.

i.e., within some 'radius of convergence' the Taylor series will converge. Let this = R . R can be determined from Ratio and Cauchy tests.

(1) Ratio: $\frac{1}{R} = \lim_{n \rightarrow \infty} \left| \frac{a_{n+1}}{a_n} \right|$ if limit exists.

(2) Cauchy: $\frac{1}{R} = \lim_{n \rightarrow \infty} |a_n|^{1/n}$

Prog results from convergence properties of tests and desired properties of R being satisfied when the above are made as an hypothesis.

Note: A function $f(z)$ is ANALYTIC at $z = z_0$ if it has a Taylor expansion about $z = z_0$ with a non-zero radius of convergence.

i.e., for $f(z) = \sum_{n=0}^{\infty} a_n (z - z_0)^n$; $\frac{1}{\lim_{n \rightarrow \infty} |a_n|^{1/n}} \neq 0$.
or use ratio test.

2nd order ODE's. (linear)

The general form of a 2nd order linear ODE is: (in $y(x)$):

$$y'' + p(x)y' + q(x)y = f(x)$$

It is homogeneous if $f(x) = 0$ and inhomogeneous if $f(x) \neq 0$

The general solution will be of the form

$$y(x) = y_c(x) + y_p(x)$$

where $y_c(x)$ is the solution to the homogeneous case and $y_p(x)$ is another solution to the ODE.

$y_c(x)$ must be composed of two linearly independent functions s.t.

$$y_c = c_1 y_1 + c_2 y_2$$

where c_1, c_2 are arbitrary constants defined by the b.c.'s of the ODE (if applicable).

Linear independence is proved by computing the Wronskian:

$$W(x) = \begin{vmatrix} y_1 & y_2 \\ y_1' & y_2' \end{vmatrix} \neq 0 \text{ if } y_1, y_2 \text{ linearly independent.}$$

Solution methods: Homogeneous case:

(1) Series solution. Try solution $y(x) = \sum_{n=0}^{\infty} a_n x^n$

This will work if ODE is

analytic at $x=0$ i.e. $x=0$ is an ORDINARY POINT. $\rightarrow$ can try any Taylor series about

ordinary point $x=x_0$ by modifying $x \rightarrow x-x_0$

If $x=0$ is not analytic $\Rightarrow$ singular point (i.e. ∞ crops up). If $p(x) = x$ and $q(x) = x^2$ are analytic

at $x=0$ then $x=0$ is a regular singular point.

Otherwise it is an irregular or essential singularity and it is not that helpful.

For regular singular points try a power series

(Frobenius series) $y(x) = x^\sigma \sum_{n=0}^{\infty} a_n x^n$

→ computing coefficients of x^0 yields a quadratic for σ . Note some solutions of this (when σ differs by an integer) will NOT work. ONE solution will always work though → use Wronskian to find the second solution.

Idea is to substitute power series into ODE and change indexes of x to yield a recurrence relation for a_n . Use this and linear independence considerations to find $y_c = c_1 y_1 + c_2 y_2$. (i.e. find y_1 and y_2 then find c_1, c_2 via b.c.'s if any).

For values of σ which don't work: Find 1 solution then use Wronskian.

$$W(x) = y_1 y_2' - y_2 y_1' \Rightarrow \frac{W}{y_1^2} = \frac{y_2'}{y_1} - \frac{y_2 y_1'}{y_1^2}$$

$$= \frac{d}{dx} \left(\frac{y_2}{y_1} \right) \quad \left[\text{check: } \frac{d}{dx} \left(\frac{y_2}{y_1} \right) = \frac{y_1 y_2' - y_2 y_1'}{y_1^2} = \frac{y_2'}{y_1} - \frac{y_2 y_1'}{y_1^2} \right]$$

$$W' = y_1 y_2'' - y_1'' y_2$$

↑ use fact that y_1, y_2 satisfy ODE to yield

$$W' = -pW \quad \int_{x_0}^x p(x') dx'$$

$$\Rightarrow W = C e^{-\int_{x_0}^x p(x') dx'}$$

$$\Rightarrow \text{us } W = \left(\exp \left[\int_{x_0}^x -p(x') dx' \right] \right)$$

(we can make $C=1$ as it can be 'absorbed' by c_1, c_2)

$$= \exp \left[\int_{x_0}^x -p(x') dx' \right] \quad \uparrow \text{dummy variable}$$

$$\Rightarrow \int \frac{W}{y_1^2} dx' = \frac{y_2}{y_1} \Rightarrow y_2 = y_1(x) \int \frac{1}{y_1^2(x')} \exp \left[\int_{x_0}^{x'} p(x') dx' \right] dx'$$

so if y_1 is known we can find y_2 .

OTHERWISE Try a substitution and try to find y_c analytically (method of series method).

For Inhomogeneous cases : $f(x) \neq 0$.

(2) Method of variation of parameters For $y_i = c_1 y_1(x) + c_2 y_2(x)$

Let $y_p(x) = \underbrace{c_1(x)}_{\text{function}} y_1(x) + \underbrace{c_2(x)}_{\text{function}} y_2(x)$.
↑ constants known.

and restrict :
$$\begin{cases} c_1' y_1 + c_2' y_2' = f \\ c_1' y_1 + c_2' y_2 = 0 \end{cases}$$

Allows c_1, c_2 to be found. Integration !
yields y_p .

(3) Green's functions.

2nd order, linear

Define $G(x, z)$ s.t. for a given ODE of b.c.
 $y(\alpha) = 0 \quad y(\beta) = 0$ as:

$$y(x) = \int_{\alpha}^{\beta} G(x, z) f(z) dz \quad (*)$$

By substitution into the ODE it appears $G(x, z)$ has properties of $\delta(x-z)$. δ is the operation of the ODE on $G(x, z)$

name :
$$G'' + p G' + q G = \delta(x-z).$$

So for $x > z$, $x < z$ we have a homogeneous ODE to solve for G which we can integrate up using (*)

$$y(x) = \int_{\alpha}^x G(x, z) f(z) dz + \int_x^{\beta} G(x, z) f(z) dz$$

$\uparrow$ $x > z$ $\uparrow$ $x < z$

Note that the solutions of the homogeneous equation will be functions of z . These can be found by using:

$G(x, z)$ must satisfy b.c.'s of ODE. ~~Derivatives~~ up to First derivative of G at $x = z$ has unit discontinuity whereas higher derivatives are continuous.

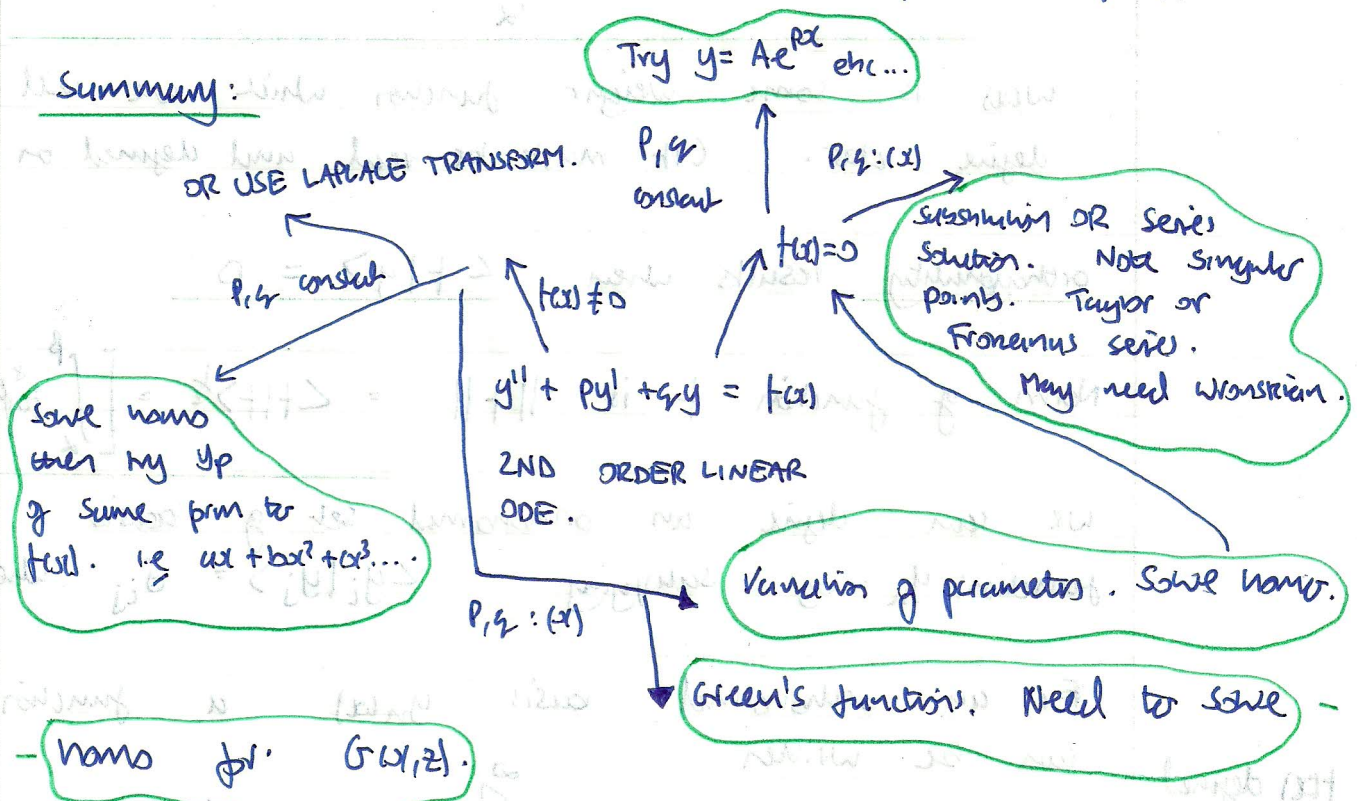
i.e. if $G(x, z) = A(z) y(x) + B(z) p(x)$

(1) $A(z) y(z) - B(z) p(z) = 0$

(2) $A(z) y'(z) - B(z) p'(z) = 1$

These conditions should allow $G(x, z)$ to be found.

Summary:



Note / b.c.'s must be homogeneous. If not substitute polynomial that fits b.c.s s.t. 'y' = y - p(x) i.e. so 'y'(x = b.c.) = 0.

Eigenfunctions and Sturm-Liouville Formulation

Analogous to a vector space we may consider a set of functions defined on an interval $[a, b]$ which, since the interval is infinite, form an infinite set of 'basis functions' which all other functions on that interval may be expressed in terms of. [Assume basis functions are linearly independent]. ~~for these~~

$$\text{i.e., } f(x) = \sum_{n=0}^{\infty} c_n y_n(x) \quad \text{where } y_n(x) = \text{basis function.}$$

Define inner product (like scalar product with matrices)

$$\langle f(x) | g(x) \rangle = \int_a^b f^*(x) g(x) w(x) dx$$

$w(x)$ is some weight function which we will define later. (It must be real and defined on $[a, b]$).

orthogonality results when $\langle f | g \rangle = 0$

$$\text{Norm of function } f \text{ is } \|f\| = \langle f | f \rangle^{1/2} = \left[\int_a^b f^* f w(x) dx \right]^{1/2}$$

We can define an orthonormal set of basis functions y_n by satisfying $\langle y_i | y_j \rangle = \delta_{ij}$ where $\|y_i\| = 1$.

For an orthonormal basis $y_n(x)$ a function $f(x)$ can be written

$$f(x) = \sum_{n=0}^{\infty} a_n y_n(x) \quad \text{where } a_n = \langle y_n | f(x) \rangle$$

This restriction on $a_n \Rightarrow$ orthonormality of basis functions a useful property.

useful inequalities

For an infinite set g functions on $[\alpha, \beta]$ with non zero inner product (A Hilbert space):

(i) Schwarz inequality $|\langle f|g \rangle| \leq \langle f|f \rangle^{1/2} \langle g|g \rangle^{1/2}$
~~where~~ when $f = \lambda g$, $\lambda \in \mathbb{C}$ equality holds.

(ii) Triangle inequality $\|f+g\| \leq \|f\| + \|g\|$
If $f = \lambda g$ equality holds. ($\lambda \in \mathbb{C}$).

(iii) Bessel's inequality: For orthonormal basis e_n :
 $\langle f|f \rangle = \|f\|^2 \geq \sum_{n=0}^{\infty} |c_n|^2$

Self Adjoint or Hermitian operators.

For a linear operator L acting on a function $y(x)$ L is self adjoint or Hermitian if

$$\langle u|Lv \rangle = \langle v|Lu \rangle^* \quad \text{For some functions } u(x), v(x).$$

operators of the form $L = -\frac{\partial}{\partial x} \left(p(x) \frac{\partial}{\partial x} \right) + q(x)$

are Hermitian. $\rightarrow$ S.L operator. We can extend matrix analogy by considering:

$$Ly = \lambda y \quad \lambda \in \mathbb{C}.$$

$y =$ 'eigenfunction' $\lambda =$ eigenvalue.

It turns out for Hermitian operators $\{\lambda\}$ are real and eigenfunctions

are orthogonal if λ are distinct.

$w(x)$ is defined to convert a non self adjoint / Hermitian $\mathcal{L}$ into one which is.

Then $\mathcal{L}y = \lambda wy$ will be a general eigenfunction equation where $\mathcal{L}$ is Hermitian $= \frac{1}{w} \mathcal{L}'$ where $\mathcal{L}$ may not be Hermitian.

or..... recognise the $\lambda w(x)$ from the equation $\mathcal{L}y = \lambda wy$ and multiply by integrating factor

$$F(x) = \exp \left[\int^x \frac{r(z) - p'(z)}{p(z)} dz \right]$$

for general $\mathcal{L}y = \lambda wy$ as

$$p(x)y'' + r(x)y' + q(x)y + \lambda w(x)y = 0$$

For self adjoint form $p'(x) = r(x)$.

so if not then multiply by $F(x)$.

use $\mathcal{L}$ S-L p.m : ① Eigenfunction expansions
② Solving inhom. ODE's via....

$$G(x,z) = \sum_{n=0}^{\infty} \frac{1}{\lambda_n} y_n(x) y_n^*(z)$$

orthonormal.

λ_n = nth eigenvalue / y_n = nth eigenfunction.

So solving $\mathcal{L}y = w(x)\lambda y$ for y_n, λ_n can be used to solve $\mathcal{L}y = f(x)$.

NOTE

what is the purpose of Eigenfunctions and S.L. theory?

with analogy to Hermitian matrix $\underline{H}$, Hermitian operators $\hat{L}$ can be present in eigenvalue equations:

$$\underline{H} \underline{y} = \lambda \underline{y} \leftrightarrow \hat{L} y = \lambda y$$

Now if $\underline{H}$ is known $\{\lambda\}$ and $\{\underline{y}\}$ can be found using

or weight function if $\hat{L}$ was not originally S. self.

$|\underline{H} - \underline{I} \lambda| = 0$ to find $\{\lambda\}$ and then back substituting to find $\{\underline{y}\}$.

This does not appear to be possible precisely $\hat{L} y = \lambda y$.
(Because order of $\hat{L}$ is $\infty + \infty$?)

$\therefore$ Since for given $\hat{L}$, y and λ cannot be explicitly determined what is the point?

$\rightarrow$ guess λ and solve $\hat{L} y = \lambda y$ try or guess y and use Rayleigh Ritz to find λ .

$\rightarrow$ For eigenfunction expansion (orthonormal eigenfunctions)

$$f(x) = \sum_{i=0}^{\infty} y_i(x) c_i \quad c_i = \langle y_i | f \rangle$$

where $f(x)$ satisfies Dirichlet conditions on boundary
 $\rightarrow$ How does one determine y_i ? $\rightarrow$ seems like one must invert a S.L. operator and solve for y_i λ_i subject to b.c.s. Wouldn't it be easier to write down a Taylor series for $f(x)$. Seems far less ambiguous!

(Though I suppose if orthogonality of 'basis' is required a S.L. approach may be helpful - However we can always use a Gram Schmidt procedure if we know our expansion is orthogonal...)

If S.L. approach is used to solve ODE's.

→ For homogeneous cases the S.L. approach seems to be: "solve the homogeneous equation (with a λ uniquely placed) and then recognise eigenfunctions". → Surely we have to already solve the ODE for this method to work!

→ For inhomogeneous cases if we solve $dy = w dy$ (i.e. for homogeneous case) we can recover a pm of $G(w, z)$. Are we not effectively solving the hom case to find $G(w, z)$ $\alpha < z$, $\alpha > z$ anyway? Again what is the point with the machinery of S.L. concept?

$$f(x) = \sum_{n=0}^{\infty} a_n x^n$$

→ How does the behaviour of $f(x)$ as $x \rightarrow \infty$ depend on the behaviour of a_n as $n \rightarrow \infty$?
 → How does the behaviour of $f(x)$ as $x \rightarrow 0$ depend on the behaviour of a_n as $n \rightarrow \infty$?
 → How does the behaviour of $f(x)$ as $x \rightarrow \infty$ depend on the behaviour of a_n as $n \rightarrow \infty$?

Calculus of Variations

Define FUNCTIONAL $F[y]$ as the integral:

$$F[y] = \int_a^b dx f(y, y'; x) \quad y' \equiv \frac{dy}{dx}$$

i.e., f is a function of y and y' which are in turn functions of independent variable x .

For a problem: "what are the functions $y(x)$ s.t. $F[y]$ is stationary" we can consider the effect on F of changing y by a small amount δy .

$$\begin{aligned} \text{i.e. } \delta F &= F[y + \delta y] - F[y] \\ &= \int_a^b dx f(y + \delta y, y' + (\delta y)'; x) - \int_a^b dx f(y, y'; x) \\ &= \int_a^b dx \left[\delta y \frac{\partial f}{\partial y} + (\delta y)' \frac{\partial f}{\partial y'} \right] + O(\delta y^2) \end{aligned}$$

As by two variable Taylor series:

$$f(y + \delta y, y' + (\delta y)'; x) \approx \delta y \frac{\partial f}{\partial y} + (\delta y)' \frac{\partial f}{\partial y'} + \underbrace{f(y, y')}_{\text{This term cancels.}}$$

So for stationary $F[y]$, $\delta F = 0$

$$\Rightarrow 0 = \int_a^b dx \left(\delta y \frac{\partial f}{\partial y} + (\delta y)' \frac{\partial f}{\partial y'} \right) \quad \text{neglecting } \delta y^2 \text{ terms.}$$

$$\Rightarrow 0 = \int_a^b dx \delta y \frac{\partial f}{\partial y} + \left[\delta y \frac{\partial f}{\partial y'} \right]_a^b - \int_a^b \delta y \frac{d}{dx} \left(\frac{\partial f}{\partial y'} \right) dx$$

$$\Rightarrow \left[\delta y \frac{\partial f}{\partial y'} \right]_a^b + \int_a^b dx \left(\delta y \frac{\partial f}{\partial y} - \delta y \frac{d}{dx} \left(\frac{\partial f}{\partial y'} \right) \right)$$

require $\delta y(b) = \delta y(a) = 0 \Rightarrow \frac{\partial f}{\partial y} - \frac{d}{dx} \left(\frac{\partial f}{\partial y'} \right) = 0$ as δy is arbitrary within $[a, b]$ for equation to hold.

yields Euler equation for F, y, y'

$$\frac{\partial F}{\partial y} = \frac{d}{dx} \left(\frac{\partial F}{\partial y'} \right)$$

So if F does not contain y explicitly:

$$\frac{\partial F}{\partial y} = 0 \Rightarrow \frac{\partial F}{\partial y'} = \text{constant.}$$

If F does not contain x explicitly:

→ multiply E eq by y' and note:

$$\frac{d}{dx} \left(y' \frac{\partial F}{\partial y'} \right) = y' \frac{d}{dx} \left(\frac{\partial F}{\partial y'} \right) + \frac{\partial F}{\partial y'} y''$$

$$\Rightarrow y' \frac{\partial F}{\partial y} = y' \frac{d}{dx} \left(\frac{\partial F}{\partial y'} \right) \Rightarrow y'' \frac{\partial F}{\partial y'} + y' \frac{\partial F}{\partial y} = \frac{d}{dx} \left(y' \frac{\partial F}{\partial y'} \right)$$

So if no explicit x dependence: → LHS = $\frac{d}{dx} F$

$$\text{i.e., } dF = dy \frac{\partial F}{\partial y} + dy' \frac{\partial F}{\partial y'} \quad \text{by chain rule}$$

$$(dy' = dy') \quad \text{hence} \quad F = y' \frac{\partial F}{\partial y'} + C.$$

Use these relations to solve 'variational' problems.

Solution method: Express problem in Functional form.

$$F[y] = \int_a^b dx f(y, y', x)$$

on some interval $[a, b]$. Apply Euler eq and use simplification where appropriate. (i.e. inspect integrand of functional).

If functional is unconstrained by another functional = constant

i.e.,
$$J = \int_a^b y(x, y, y') dx = \text{constant}$$

Define new functional $K = F + \lambda J$ and extremise this. (i.e. find λ via constraint - method of Lagrange multipliers).

→ i.e., Lagrange's equation of motion
Fermat's = principle in optics.

use this method.

S.L. problems and calculus of variations

Considering the functional $F[y] = \int_a^b [p(x)y'^2 + q(x)y^2] dx$
for $p \neq 0$ for $a < x < b$

constrained by $G[y] = 1 = \int_a^b w(x)y^2 dx$ where $w \neq 0$ for $a < x < b$

→ Minimising $\Rightarrow \delta F - \lambda \delta G = 0$ (L. multipliers).

↓ Applying Euler equation to new functional $F - \lambda G$
yields S.L. equation

$$-(py')' + qy = \lambda wy. \Leftrightarrow \lambda y = 2wy.$$

i.e. λ is the eigenvalue of the S.L. equation.

$$\text{let } \lambda = \frac{F}{G} \Rightarrow \delta \lambda = \frac{G \delta F - F \delta G}{G^2} = \frac{1}{G} (\delta F - \lambda \delta G)$$

So for stationary λ , $\delta F - \lambda \delta G = 0$ i.e. $\lambda = 2$.

(we can do this as $G = \text{constant}$ so extremising Λ is equivalent to extremising F).

So when Λ is stationary:

$$\Lambda = \frac{\int_a^b (\rho y'^2 + \sigma y^2) dx}{\int_a^b w y^2 dx}$$

where y = eigenfunction i.e. solution to S.L. equation. If we guess then we can $\therefore$ guess Λ by this method.

If we guess with an eigenfunction of 'adjustable' parameters we can find Λ and then minimise w.r.t. this parameter to find minimum or stationary Λ .

If we include more than one adjustable parameter minimise or eigenvalue by considering $\nabla \Lambda = 0$. ~~That 'vector space'~~ of Imagin Λ as a function of parameters

Fin!