Differential Equations

$$y' + a(x)y = b(x) \longrightarrow h(x) = e^{\int a(x) dx}$$
2 distinct λ : $y_H = c_1 e^{\lambda_1 x} + c_2 e^{\lambda_2 x}$
1 repeated λ : $y_H = c_1 x e^{\lambda x} + c_2 e^{\lambda x}$

$$\lambda = \alpha \pm \omega i : y_H = e^{\alpha x} \left(c_1 \cos \omega x + c_2 \sin \omega x \right).$$

$$k e^{\alpha x} \longrightarrow c e^{\alpha x}$$

$$k x^n \longrightarrow \sum_{i=0}^n c_i x^i$$

$$k \cos \alpha x$$
 or $k \sin \alpha x \longrightarrow c_1 \cos \alpha x + c_2 \sin \alpha x$
 $(\cdots) e^{\alpha x} \longrightarrow (\cdots) e^{\alpha x}$

$$W(y_1, y_2)(x) = \det \begin{pmatrix} y_1 & y_2 \\ y'_1 & y'_2 \end{pmatrix} = y_1 y'_2 - y_2 y'_1$$

$$r = y'' + py' + qy y_P = uy_1 + vy_2$$

$$u = -\int \frac{y_2 r}{W} dx v = \int \frac{y_1 r}{W} dx$$

$$\cosh x = \frac{e^x + e^{-x}}{2} \qquad \qquad \sinh x = \frac{e^x - e^{-x}}{2}$$
$$\cosh ix = \cos x \qquad \qquad \sinh ix = i \sin x$$

Projections and Orthonormal Bases

$$P_{\beta' \to \beta''} P_{\beta \to \beta'} = P_{\beta \to \beta''}$$

$$\|\mathbf{v}\| = \sqrt{\langle \mathbf{v}, \mathbf{v} \rangle}$$

$$d(\mathbf{u}, \mathbf{v}) = \|\mathbf{u} - \mathbf{v}\|$$

$$|\langle \mathbf{u}, \mathbf{v} \rangle| \le \|\mathbf{u}\| \|\mathbf{v}\|$$

$$\|\mathbf{u} + \mathbf{v}\| \le \|\mathbf{u}\| + \|\mathbf{v}\|$$

$$\|\mathbf{u} + \mathbf{v}\|^2 = \|\mathbf{u}\|^2 + \|\mathbf{v}\|^2 \iff \langle \mathbf{u}, \mathbf{v} \rangle = 0$$

$$U^{\perp} = \{\mathbf{v} \in V \mid \langle \mathbf{v}, \mathbf{u} \rangle = 0 \ \forall \mathbf{u} \in U\}$$

$$\operatorname{Proj}_{U}(\mathbf{v}) = \langle \mathbf{v}_{1}, \hat{e}_{1} \rangle \hat{e}_{1} + \dots + \langle \mathbf{v}, \hat{e}_{k} \rangle \hat{e}_{k}$$

$$\operatorname{Proj}_{U_{\perp}}(\mathbf{v}) = \mathbf{v} - \operatorname{Proj}_{U}(\mathbf{v})$$

$$A\mathbf{x} = \mathbf{b} \longrightarrow A^{\mathrm{T}} A\mathbf{x} = A^{\mathrm{T}} \mathbf{b} \longrightarrow \mathbf{x} = (A^{\mathrm{T}} A)^{-1} A^{\mathrm{T}} \mathbf{b}$$

Matrix-Related Computation

$$A\mathbf{x} = \lambda \mathbf{x}$$
 $\det(A - \lambda I) = 0$ $(A - \lambda I)\mathbf{x} = \mathbf{0}$

AP = PD, where $A = PDP^{-1}$ for diagonalisation and $A = PDP^{T}$ for orthogonal diagonalisation.

$$P = \begin{bmatrix} \mathbf{v_1} & \mathbf{v_2} & \cdots \end{bmatrix} \quad D = \begin{pmatrix} \lambda_1 & 0 & 0 \\ 0 & \ddots & 0 \\ 0 & 0 & \lambda_n \end{pmatrix}$$

P's vectors need to be *orthonormal* when orthogonally diagonalising.

$$ax^{2} + by^{2} + cxy \longrightarrow \begin{pmatrix} a & c/2 \\ c/2 & b \end{pmatrix}$$

$$ax^{2} + by^{2} + cz^{2} + dxy \longrightarrow \begin{pmatrix} a & d/2 & e/2 \\ d/2 & b & f/2 \\ e/2 & f/2 & c \end{pmatrix}$$

$$\mathbf{x}^{\mathrm{T}}A\mathbf{x} + K\mathbf{x} + c = 0$$

When orthogonally diagonalising coefficient matrix of quadratic form, arrange column vectors of P such that $\det P = +1$.

Taylor Series and Critical Points

$$H_f = \begin{pmatrix} f_{x_{1x_1}} & \cdots & f_{x_{1}x_n} \\ \vdots & \ddots & \vdots \\ f_{x_nx_1} & \cdots & f_{x_nx_n} \end{pmatrix}$$

$$f(\mathbf{x}) \approx f(\mathbf{x_0}) + (\nabla f(\mathbf{x_0}))^{\mathrm{T}} (\mathbf{x} - \mathbf{x_0})$$
$$+ \frac{1}{2} (\mathbf{x} - \mathbf{x_0})^{\mathrm{T}} H_f(\mathbf{x_0}) (\mathbf{x} - \mathbf{x_0})^{\mathrm{T}} + \cdots$$

$$f(\mathbf{x} + \mathbf{h}) \approx \sum_{l=0}^{\infty} \frac{1}{l!} (\mathbf{h} \cdot \nabla)^l f(\mathbf{x})$$

• Minimum: all $\lambda > 0$

• Maximum: all $\lambda < 0$

• Saddle: there are λ with different signs

• Inconclusive: there are some λ which equal 0 and the rest (non-zero λ) have the same sign

Double and Triple Integrals

For $a, b, c, d \in \mathbb{R}$,

$$\int_{a}^{b} \int_{c}^{d} f(x)g(y) \, dy \, dx = \left(\int_{a}^{b} f(x) \, dx\right) \left(\int_{c}^{d} g(y) \, dy\right)$$

$$\begin{aligned} &\text{area} = \iint_D 1 \, \mathrm{d}A & \text{average} = \frac{\iint_D f(x,y) \, \mathrm{d}A}{\iint_D 1 \, \mathrm{d}A} \\ &\text{volume} = \iiint_V 1 \, \mathrm{d}V & \text{average} = \frac{\iiint_V f(x,y,z) \, \mathrm{d}V}{\iiint_V 1 \, \mathrm{d}v} \end{aligned}$$

$$m = \iiint_V \rho(x, y, z) \, dV \qquad M_{yz} = \iiint_V x \rho(x, y, z) \, dV$$
$$M_{xz} = \iiint_V y \rho(x, y, z) \, dV \qquad M_{xy} = \iiint_V z \rho(x, y, z) \, dV$$

$$\bar{x} = \frac{M_{yz}}{m}$$
 $\bar{y} = \frac{M_{xz}}{m}$ $\bar{z} = \frac{M_{xy}}{m}$

For 2D, use $\iint_A (\cdots) dA$ with $\rho(x, y)$ and z = 0. Then use $M_{yz} = M_y$, $M_{xz} = M_x$, $M_{xy} = 0$.

Coordinate Systems

Polar	$x = r\cos\theta \qquad y = r\sin\theta$
	$r^2 = x^2 + y^2 J = r$
Cylindrical	Polar but with z parameter
	J = r
Spherical	$x = r\cos\theta\sin\varphi y = r\sin\theta\sin\varphi$
	$z = r\cos\varphi \qquad \qquad r^2 = x^2 + y^2 + z^2$
	$ J = r^2 \sin \varphi$

$$\text{Jacobian} = |J| = \left| \det \begin{pmatrix} \frac{\partial x}{\partial u} & \frac{\partial x}{\partial v} & \frac{\partial x}{\partial w} \\ \frac{\partial y}{\partial u} & \frac{\partial y}{\partial v} & \frac{\partial y}{\partial w} \\ \frac{\partial z}{\partial u} & \frac{\partial z}{\partial v} & \frac{\partial z}{\partial w} \end{pmatrix} \right|$$

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Vector Calculus

$$\nabla = \left(\frac{\partial}{\partial x}, \frac{\partial}{\partial y}, \frac{\partial}{\partial z}\right)$$

$$D_{\hat{\mathbf{u}}}(f) = (\nabla f) \cdot \hat{\mathbf{u}}$$

$$W = \int_{C} \mathbf{F} \cdot d\mathbf{r} = \int_{a}^{b} \mathbf{F}(\mathbf{r}(t)) \cdot \mathbf{r}'(t) dt$$

$$\int_{C} \nabla f \cdot d\mathbf{r} = f(\mathbf{r}(b)) - f(\mathbf{r}(a))$$

$$\operatorname{div} \mathbf{v} = \nabla \cdot \mathbf{v} = \frac{\partial v_{1}}{\partial x} + \frac{\partial v_{2}}{\partial y} + \frac{\partial v_{3}}{\partial z}$$

$$\operatorname{curl} \mathbf{v} = \nabla \times \mathbf{v} = \begin{vmatrix} \hat{\mathbf{i}} & \hat{\mathbf{j}} & \hat{\mathbf{k}} \\ \frac{\partial}{\partial x} & \frac{\partial}{\partial y} & \frac{\partial}{\partial z} \\ v_{1} & v_{2} & v_{3} \end{vmatrix}$$

$$(\mathbf{r}_{\mathbf{u}}(a, b) \times \mathbf{r}_{\mathbf{v}}(a, b)) \cdot ((x, y, z) - \mathbf{r}(a, b)) = 0$$

$$\iint_{S} f(x, y, z) dS = \iint_{D} f(\mathbf{r}(u, v)) \|\mathbf{r}_{\mathbf{u}} \times \mathbf{r}_{\mathbf{v}}\| dA$$

Miscellaneous

$$\int \frac{1}{\sqrt{x^2 + 1}} dx = \operatorname{arsinh}(x) + c$$

$$\int \frac{1}{\sqrt{x^2 - 1}} dx = \operatorname{arcosh}(x) + c, x > 1$$

$$\operatorname{arsinh}(x) = \ln\left(x + \sqrt{x^2 + 1}\right)$$

$$\operatorname{arcosh}(x) = \ln\left(x + \sqrt{x^2 - 1}\right), x \ge 1$$

$$\operatorname{artanh}(x) = \frac{1}{2}\ln\left(\frac{1 + x}{1 - x}\right), x \in (-1, 1)$$

For some $A \in M_{n \times n}(\mathbb{F})$, the following statements are equivalent:

- A is non-singular $(A^{-1} \text{ exists})$
- Only $\mathbf{x} = \mathbf{0}$ satisfies $A\mathbf{x} = \mathbf{0}$
- The row-echelon form of A does not have a row of zeroes
- $A\mathbf{x} = \mathbf{b}$ has a solution for all $\mathbf{b} \in \mathbb{R}^n$
- $\det A \neq 0$
- Columns of A are linearly independent
- Rows of A are linearly independent
- $\dim(NS(A)) = 0$
- $\operatorname{rank} A = n$
- $\lambda = 0$ is not an eigenvalue of A

$$\begin{aligned} &\text{flux} = \int_{C} \mathbf{v} \cdot \mathbf{n} \, \mathrm{d}S = \int_{a}^{b} \mathbf{v}(\mathbf{r}(t)) \cdot \underbrace{\left(\mathbf{r}'(t) \times \hat{\mathbf{k}}\right)}_{\text{check orientation}} \, \mathrm{d}t \\ &\text{flux} = \iint_{S} \mathbf{v} \cdot \mathbf{n} \, \mathrm{d}S = \iint_{D} \mathbf{v} \cdot \left(\mathbf{r}_{\mathbf{u}} \times \mathbf{r}_{\mathbf{v}}\right) \, \mathrm{d}A \end{aligned}$$

$$\iint_{D} \frac{\partial F_{2}}{\partial x} - \frac{\partial F_{1}}{\partial y} dA = \oint_{\partial D} \mathbf{F} \cdot d\mathbf{r}$$

$$\oint_{\partial D} \mathbf{v}(x, y) \cdot \mathbf{n} dS = \iint_{D} \mathbf{\nabla} \cdot \mathbf{v}(x, y) dA$$

$$\oiint_{S} \mathbf{F} \cdot \mathbf{n} dS = \iiint_{V} \mathbf{\nabla} \cdot \mathbf{F} dV$$

$$\iint_{S} (\mathbf{\nabla} \times \mathbf{F}) \cdot \mathbf{n} dS = \oint_{\partial S} \mathbf{F} \cdot d\mathbf{r}$$

 $A \in M_{n \times n}(\mathbb{F})$ is diagonalisable $(A = PDP^{-1})$

iff A has n linearly independent eigenvectors

iff algebraic and geometric multiplicities are equal for every eigenvalue

 $A \in M_{n \times n}(\mathbb{R})$ is orthogonally diagonalisable $(A = PDP^{\mathrm{T}})$

iff A is symmetric, i.e. $A=A^{\rm T}$

iff eigenvectors corresponding to different eigenvalues are orthogonal with respect to the dot product

The complex inner product for $\mathbf{u}, \mathbf{v} \in \mathbb{C}^n$ is defined by

$$\mathbf{u} \cdot \mathbf{v} = u_1 \bar{v}_1 + \dots + u_n \bar{v}_n = \mathbf{v}^* \mathbf{u} \in \mathbb{C}$$

where \bar{z} is the conjugate of z and \mathbf{v}^* is the conjugate transpose of \mathbf{v} . Now, $\overline{\langle \mathbf{u}, \mathbf{v} \rangle} = \langle \mathbf{v}, \mathbf{u} \rangle$.