

Differential Equations

$$y' + a(x)y = b(x) \longrightarrow h(x) = e^{\int a(x) dx}$$

$$2 \text{ distinct } \lambda: y_H = c_1 e^{\lambda_1 x} + c_2 e^{\lambda_2 x}$$

$$1 \text{ repeated } \lambda: y_H = c_1 x e^{\lambda x} + c_2 e^{\lambda x}$$

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

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

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

$$k \cos \alpha x \text{ or } k \sin \alpha x \longrightarrow c_1 \cos \alpha x + c_2 \sin \alpha x$$

$$(\dots) e^{\alpha x} \longrightarrow (\dots) e^{\alpha x}$$

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

$$r = y'' + p y' + q y \quad y_P = u y_1 + v y_2$$

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

$$\cosh x = \frac{e^x + e^{-x}}{2} \quad \sinh x = \frac{e^x - e^{-x}}{2}$$

$$\cosh ix = \cos x \quad \sinh ix = i \sin x$$

Projections and Orthonormal Bases

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

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

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

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

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

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

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

$$\text{Proj}_U(\mathbf{v}) = \langle \mathbf{v}_1, \hat{e}_1 \rangle \hat{e}_1 + \dots + \langle \mathbf{v}, \hat{e}_k \rangle \hat{e}_k$$

$$\text{Proj}_{U^\perp}(\mathbf{v}) = \mathbf{v} - \text{Proj}_U(\mathbf{v})$$

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

Matrix-Related Computation

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

$\mathbf{A}\mathbf{P} = \mathbf{P}\mathbf{D}$, where $\mathbf{A} = \mathbf{P}\mathbf{D}\mathbf{P}^{-1}$ for diagonalisation and $\mathbf{A} = \mathbf{P}\mathbf{D}\mathbf{P}^T$ for orthogonal diagonalisation.

$$\mathbf{P} = [\mathbf{v}_1 \quad \mathbf{v}_2 \quad \dots] \quad \mathbf{D} = \begin{pmatrix} \lambda_1 & 0 & 0 \\ 0 & \ddots & 0 \\ 0 & 0 & \lambda_n \end{pmatrix}$$

\mathbf{P} 's vectors need to be *orthonormal* when orthogonally diagonalising.

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

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

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

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

Taylor Series and Critical Points

$$H_f = \begin{pmatrix} f_{x_1 x_1} & \dots & f_{x_1 x_n} \\ \vdots & \ddots & \vdots \\ f_{x_n x_1} & \dots & f_{x_n x_n} \end{pmatrix}$$

$$f(\mathbf{x}) \approx f(\mathbf{x}_0) + (\nabla f(\mathbf{x}_0))^T (\mathbf{x} - \mathbf{x}_0) + \frac{1}{2} (\mathbf{x} - \mathbf{x}_0)^T H_f(\mathbf{x}_0) (\mathbf{x} - \mathbf{x}_0)^T + \dots$$

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

- Minimum: $\lambda_i > 0 \forall i$
- Maximum: $\lambda_i < 0 \forall i$
- Saddle: there are λ with different signs
- Inconclusive: there exist $\lambda_i = 0$ and non-zero λ have same sign

Double and Triple Integrals

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

$$\int_a^b \int_c^d f(x)g(y) dy dx = \left(\int_a^b f(x) dx \right) \left(\int_c^d g(y) dy \right)$$

$$\text{area} = \iint_D 1 dA \quad \text{average} = \frac{\iint_D f(x, y) dA}{\iint_D 1 dA}$$

$$\text{volume} = \iiint_V 1 dV \quad \text{average} = \frac{\iiint_V f(x, y, z) dV}{\iiint_V 1 dV}$$

2D case:

2D case

$$m = \iint_D \rho(x, y) dA$$

$$M_y = \iint_D x \rho(x, y) dA$$

$$M_x = \iint_D y \rho(x, y) dA$$

3D case:

$$m = \iiint_V \rho(x, y, z) dV$$

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

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

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

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

For 2D, $M_{yz} = M_y$, $M_{xz} = M_x$, $M_{xy} = 0$.

Coordinate Systems

Polar: $x = r \cos \theta$, $y = r \sin \theta$ so $r^2 = x^2 + y^2$ and $|J| = r$.

Cylindrical: same as polar but with $z = z$, $|J| = r$.

Spherical: $x = r \cos \theta \sin \varphi$, $y = r \sin \theta \sin \varphi$, $z = r \cos \varphi$ so $r^2 = x^2 + y^2 + z^2$, $x^2 + y^2 = r^2 \sin^2 \varphi$ and $|J| = r^2 \sin \varphi$. θ is angle along x - y plane, φ is angle from point to z -axis.

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

duffy's MATH2001 cheat sheet

Vector Calculus

$$\begin{aligned}\nabla &= \left(\frac{\partial}{\partial x}, \frac{\partial}{\partial y}, \frac{\partial}{\partial z} \right) \\ D_{\hat{\mathbf{u}}}(f) &= (\nabla f) \cdot \hat{\mathbf{u}} \\ W &= \int_C \mathbf{F} \cdot d\mathbf{r} = \int_a^b \mathbf{F}(\mathbf{r}(t)) \cdot \mathbf{r}'(t) dt \\ \int_C \nabla f \cdot d\mathbf{r} &= f(\mathbf{r}(b)) - f(\mathbf{r}(a)) \\ \iint_D \frac{\partial F_2}{\partial x} - \frac{\partial F_1}{\partial y} dA &= \oint_{\partial D} \mathbf{F} \cdot d\mathbf{r} \\ \text{flux} &= \int_C \mathbf{v} \cdot \mathbf{n} dS = \int_a^b \mathbf{v}(\mathbf{r}(t)) \cdot \underbrace{(\mathbf{r}'(t) \times \hat{\mathbf{k}})}_{\text{check orientation}} dt \\ \text{div } \mathbf{v} &= \nabla \cdot \mathbf{v} = \frac{\partial v_1}{\partial x} + \frac{\partial v_2}{\partial y} + \frac{\partial v_3}{\partial z} \\ \oint_{\partial D} \mathbf{v}(x, y) \cdot \mathbf{n} dS &= \iint_D \nabla \cdot \mathbf{v}(x, y) dA \\ (\mathbf{r}_{\mathbf{u}}(a, b) \times \mathbf{r}_{\mathbf{v}}(a, b)) \cdot ((x \ y \ z) - \mathbf{r}(a, b)) &= 0 \\ \iint_S f(x, y, z) dS &= \iint_D f(\mathbf{r}(u, v)) \|\mathbf{r}_{\mathbf{u}} \times \mathbf{r}_{\mathbf{v}}\| dA \\ \text{flux} &= \iint_S \mathbf{v} \cdot \mathbf{n} dS = \iint_D \mathbf{v} \cdot (\mathbf{r}_{\mathbf{u}} \times \mathbf{r}_{\mathbf{v}}) dA \\ \oiint_S \mathbf{F} \cdot \mathbf{n} dS &= \iiint_V \nabla \cdot \mathbf{F} dV \\ \text{curl } \mathbf{v} &= \nabla \times \mathbf{v} = \begin{vmatrix} \hat{\mathbf{i}} & \hat{\mathbf{j}} & \hat{\mathbf{k}} \\ \frac{\partial}{\partial x} & \frac{\partial}{\partial y} & \frac{\partial}{\partial z} \\ v_1 & v_2 & v_3 \end{vmatrix} \\ \iint_S (\nabla \times \mathbf{F}) \cdot \mathbf{n} dS &= \oint_{\partial S} \mathbf{F} \cdot d\mathbf{r} \end{aligned}$$

this is a test $A^T A^{-T}$