

Skip §4.9: Improper Integrals

Gaussian Quadrature by Legendre polynomials

Gaussian Quadrature by Legendre polynomials: $P_0(x) = 1, P_1(x) = x,$
 $(n + 1)P_{n+1}(x) = (2n + 1)xP_n(x) - nP_{n-1}(x)$ for $n \geq 1.$

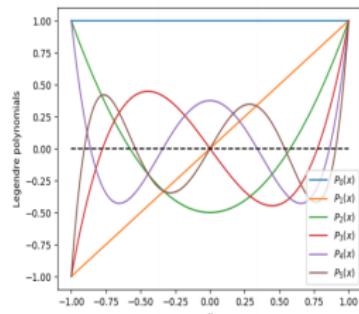
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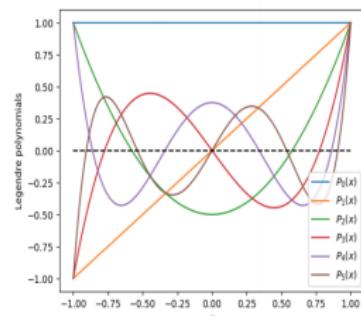


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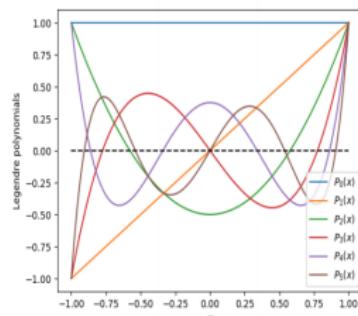
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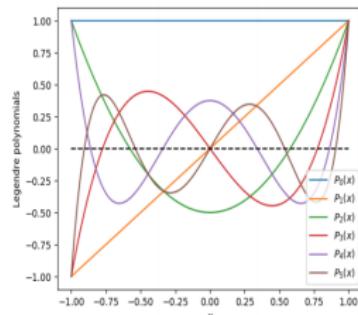
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quadrature exact for polynomial with degree $\leq n - 1$

Thm: Gaussian Quadrature $\text{DoP} = 2n - 1$

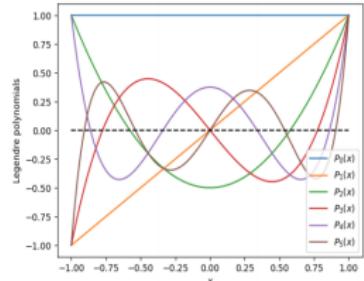
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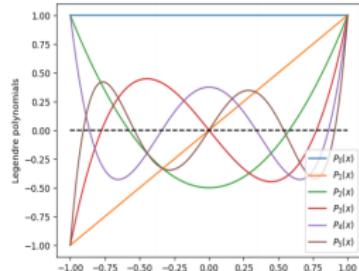


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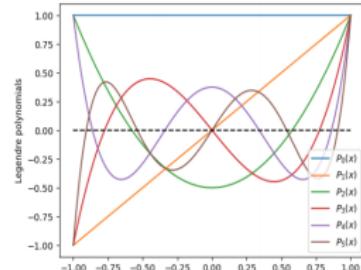
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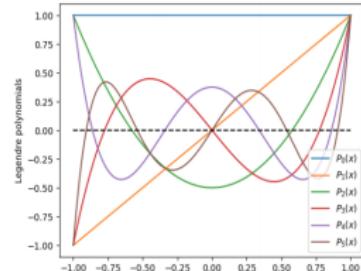
$$P(x) = P_n(x) Q(x) + R(x), \deg \text{ of } Q(x), R(x) \leq n - 1 \quad (2)$$

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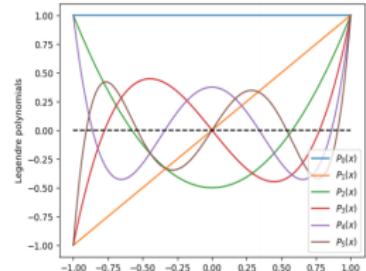
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Hermite Interpolation on roots of Legendre polynomials

- Given Legendre roots x_1, x_2, \dots, x_n with

$$(x_1, f(x_1), f'(x_1)), (x_2, f(x_2), f'(x_2)), \dots, (x_n, f(x_n), f'(x_n)),$$

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- (recall) **Thm:** For each $x \in [a, b]$, there exists $\xi(x) \in (a, b)$,

$$f(x) = H(x) + R(x), \quad R(x) = \frac{f^{(2n)}(\xi(x))}{(2n)!} (x-x_1)^2(x-x_2)^2 \cdots (x-x_n)^2.$$

Hermite Interpolation, with Gaussian Quadrature

$$\int_{-1}^1 f(x) dx = \int_{-1}^1 H(x) dx + \int_{-1}^1 R(x) dx$$

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Gaussian Quadrature Error Estimate, matlab code

$$R = \frac{f^{(2n)}(\xi)}{(2n)!} \int_{-1}^1 (x - x_1)^2(x - x_2)^2 \cdots (x - x_n)^2 dx$$

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Rapid convergence for smooth functions

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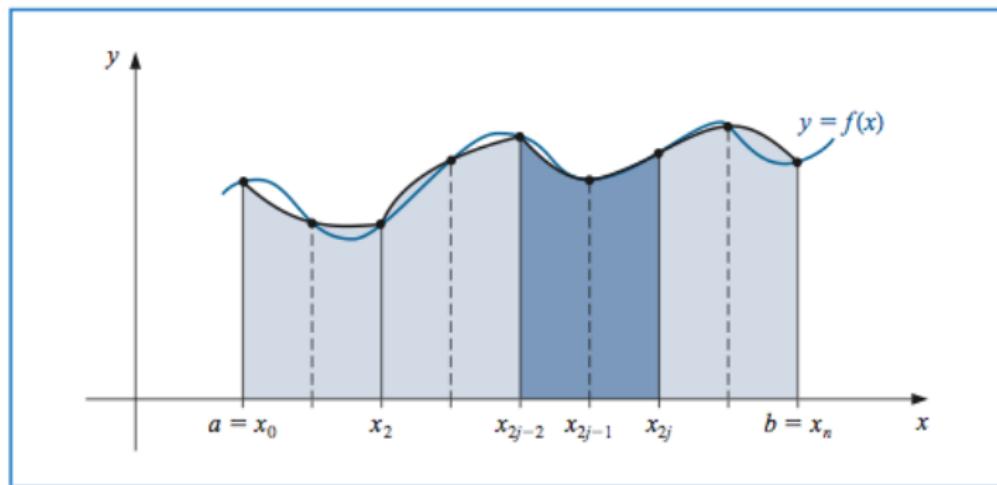
Rapid convergence for smooth functions

```
function [c,x] = Legendre(n)
%
b      = transpose((1:n-1));
b      = b./sqrt((2*b-1).*(2*b+1));
B      = diag(b,1)+diag(b,-1);
[Q,D] = eig(B);
x      = diag(D);
x(abs(x)<1e-15) = 0;
c      = 2*transpose(Q(1,:).^2);
```

In contrast: Composite Simpson's Rule

$$(n = 2m, x_j = a + j h, h = \frac{b-a}{n}, 0 \leq j \leq n)$$

$$\int_a^b f(x) dx = \frac{h}{3} \left(f(a) + 2 \sum_{j=1}^{m-1} f(x_{2j}) + 4 \sum_{j=1}^m f(x_{2j-1}) + f(b) \right) - \frac{(b-a)h^4}{180} f^{(4)}(\mu)$$



Simpson/Trapezoidal vs. Gaussian Quadratures

Simpson/Trapezoidal:

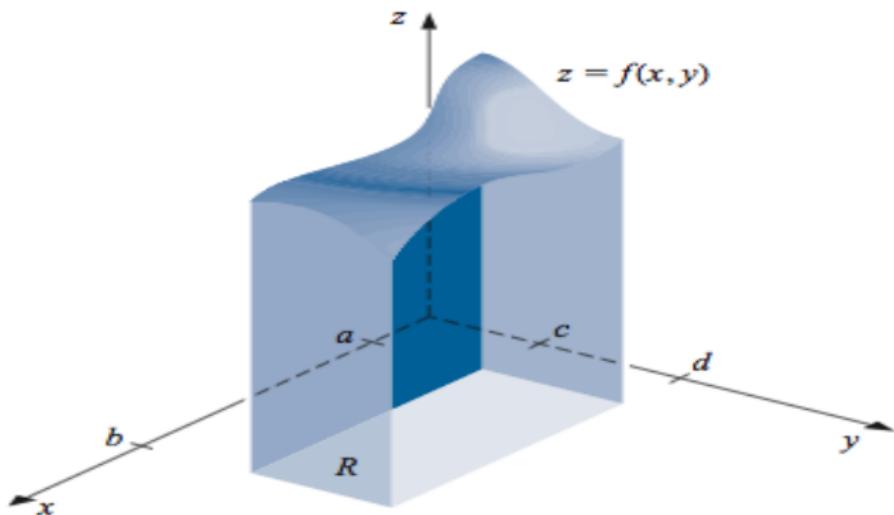
- ▶ Composite rules:
 - ▶ Adding more EQUI-SPACED points.
- ▶ Romberg extrapolation:
 - ▶ Obtaining higher order rules from lower order rules.
- ▶ Adaptive quadratures:
 - ▶ Adding more points ONLY WHEN NECESSARY.

Gaussian Quadrature:

- ▶ points different for different n .

Gaussian Quadrature good for given n ,
Simpson good for given tolerance.

§4.8 Double Integral $\int \int_R f(x, y) dA$



$$R = \{(x, y) \mid a \leq x \leq b, c \leq y \leq d\}$$

Double Integral = Integral of Integral function

$$\begin{aligned}\int \int_R f(x, y) dA &= \int_a^b \left(\int_c^d f(x, y) dy \right) dx \\ &= \int_a^b g(x) dx, \text{ with } g(x) \stackrel{\text{def}}{=} \int_c^d f(x, y) dy\end{aligned}$$

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General approach in a nutshell

- ▶ Approximate $\int_a^b g(x) dx$ with ANY quadrature

$$\int_a^b g(x) dx \approx c_1 g(x_1) + c_2 g(x_2) + \cdots + c_n g(x_n)$$

- ▶ For each node x_i , approximate $g(x_i) = \int_c^d f(x_i, y) dy$ with ANY quadrature.
- ▶ Rest is book keeping: work out 2D quadrature / error estimate

$$\int \int_R f(x, y) dA = \int_a^b g(x) dx, \quad g(x) = \int_c^d f(x, y) dy$$

► Approximate $\int_a^b g(x) dx$ with n -point quadrature:

$$\int_a^b g(x) dx = c_1 g(x_1) + c_2 g(x_2) + \cdots + c_n g(x_n) + \mathbf{R}(g)$$

$$\int \int_R f(x, y) dA = \int_a^b g(x) dx, \quad g(x) = \int_c^d f(x, y) dy$$

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► For $1 \leq i \leq n$, approximate $g(x_i)$ with m -point quadrature:

$$\int_c^d f(x_i, y) dy = \widehat{c}_1 f(x_i, y_1) + \widehat{c}_2 f(x_i, y_2) + \cdots + \widehat{c}_m f(x_i, y_m) + \widehat{\mathbf{R}}(f(x_i, \cdot)).$$

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$$\begin{aligned} \int \int_R f(x, y) dA &= \left(\sum_{i=1}^n c_i g(x_i) \right) + \mathbf{R}(g) \\ &= \left(\sum_{i=1}^n c_i \left(\left(\sum_{j=1}^m \widehat{c}_j f(x_i, y_j) \right) + \widehat{\mathbf{R}}(f(x_i, \cdot)) \right) \right) + \mathbf{R}(g) \end{aligned}$$

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$$\int \int_R f(x, y) dA = \int_a^b g(x) dx, \quad g(x) = \int_c^d f(x, y) dy$$

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$$\int_a^b g(x) dx = c_1 g(x_1) + c_2 g(x_2) + \cdots + c_n g(x_n) + \mathbf{R}(g)$$

► For $1 \leq i \leq n$, approximate $g(x_i)$ with m -point quadrature:

$$\int_c^d f(x_i, y) dy = \widehat{c}_1 f(x_1, y_1) + \widehat{c}_2 f(x_i, y_2) + \cdots + \widehat{c}_m f(x_i, y_m) + \widehat{\mathbf{R}}(f(x_i, \cdot)).$$

$$\begin{aligned} \int \int_R f(x, y) dA &= \left(\sum_{i=1}^n c_i g(x_i) \right) + \mathbf{R}(g) \\ &= \left(\sum_{i=1}^n c_i \left(\left(\sum_{j=1}^m \widehat{c}_j f(x_i, y_j) \right) + \widehat{\mathbf{R}}(f(x_i, \cdot)) \right) \right) + \mathbf{R}(g) \\ &= \boxed{\sum_{i=1}^n \sum_{j=1}^m c_i \widehat{c}_j f(x_i, y_j)} + \boxed{\left(\sum_{i=1}^n c_i \widehat{\mathbf{R}}(f(x_i, \cdot)) \right) + \mathbf{R}(g)} \\ &= \boxed{2D \text{ quadrature}} + \boxed{\text{error estimate}} \end{aligned}$$

$$\int \int_R f(x, y) dA = \sum_{i=1}^n \sum_{j=1}^m c_i \widehat{c}_j f(x_i, y_j) + \sum_{i=1}^n c_i \widehat{\mathbf{R}}(f(x_i, \cdot)) + \mathbf{R}(g)$$

$$R = \{(x, y) \mid a \leq x \leq b, c \leq y \leq d.\}$$

Example, Simpson's Rule with $m = n = 3$:

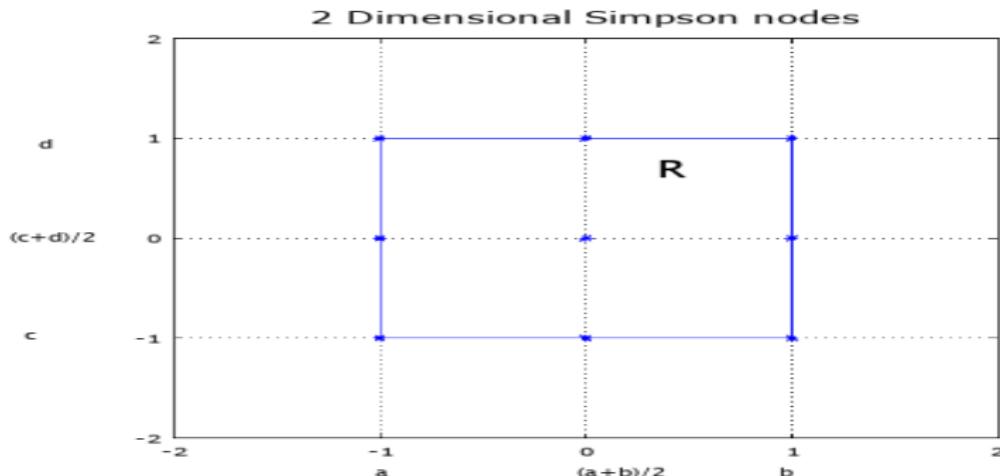
- ▶ Simpson's Rule on $[a, b]$: $(x_1, x_2, x_3) = (a, \frac{a+b}{2}, b)$.
- ▶ Simpson's Rule on $[c, d]$: $(y_1, y_2, y_3) = (c, \frac{c+d}{2}, d)$.

$$\int \int_R f(x, y) dA = \sum_{i=1}^n \sum_{j=1}^m c_i \widehat{c_j} f(x_i, y_j) + \sum_{i=1}^n c_i \widehat{\mathbf{R}}(f(x_i, \cdot)) + \mathbf{R}(g)$$

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Example, $m = n = 3$, $g(x) \stackrel{\text{def}}{=} \int_c^d f(x, y) dy$

► Simpson's Rule on $[a, b]$: $(x_1, x_2, x_3) = (a, \frac{a+b}{2}, b)$.

$$\int_a^b g(x) dx = c_1 g(x_1) + c_2 g(x_2) + c_3 g(x_3) - \frac{h^5}{90} g^{(4)}(\xi),$$

$$h = \frac{b-a}{2}, \quad (c_1, c_2, c_3) = \frac{h}{3} (1, 4, 1).$$

► Simpson's Rule on $[c, d]$: $(y_1, y_2, y_3) = (c, \frac{c+d}{2}, d)$.

$$g(x_i) = \int_c^d f(x_i, y) dy = \widehat{c}_1 f(x_i, y_1) + \widehat{c}_2 f(x_i, y_2) + \widehat{c}_3 f(x_i, y_3) - \frac{k^5}{90} \frac{\partial^4 f}{\partial^4 y}(x_i, \eta_i),$$

$$k = \frac{d-c}{2}, \quad (\widehat{c}_1, \widehat{c}_2, \widehat{c}_3) = \frac{k}{3} (1, 4, 1).$$

Composite Simpson Rules, in general

$$\int \int_R f(x, y) dA = \left(\sum_{i=1}^n \sum_{j=1}^m c_i \widehat{c}_j f(x_i, y_j) \right) - \frac{k^4 (d - c)}{180} \left(\sum_{i=1}^m c_i \frac{\partial^4 f}{\partial^4 y}(x_i, \eta_i) \right) - \frac{h^4 (b - a)}{180} \int_c^d \frac{\partial^4 f}{\partial^4 x}(\xi, y) dy$$

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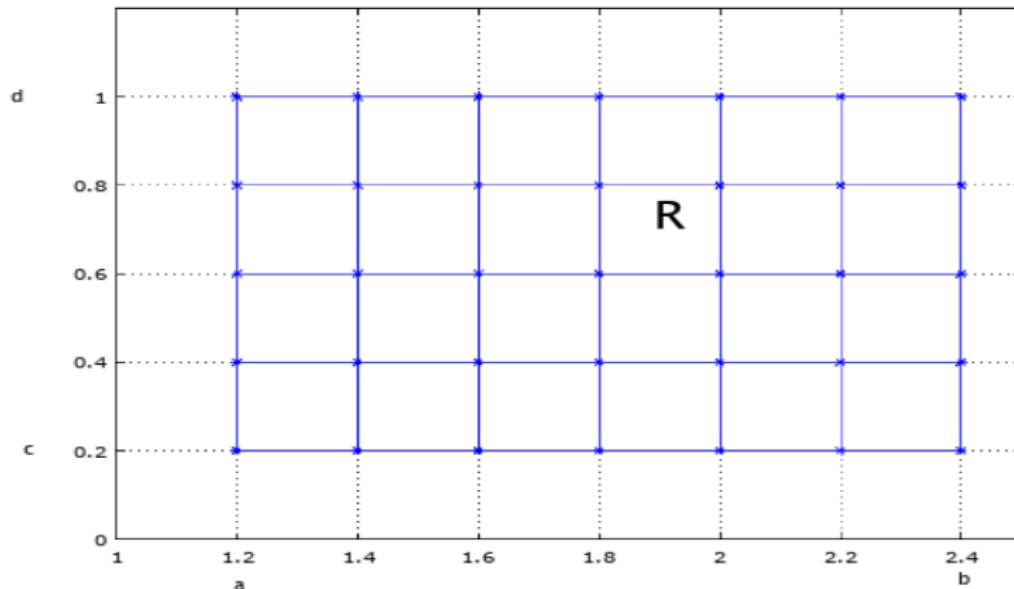
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Example: $\int \int_R \log(x + 2y) dA$ with $n = 7, m = 5$

$$R = \{(x, y) \mid 1.2 \leq x \leq 2.4, \quad 0.2 \leq y \leq 1.\}$$

Simpson Rule for double integral, $n = 7, m = 5$



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$$\begin{aligned}R &= \{(x, y) \mid 1.2 \leq x \leq 2.4, \quad 0.2 \leq y \leq 1, \} \\h &= \frac{2.4 - 1.2}{7 - 1} = 0.2, \quad k = \frac{1 - 0.2}{5 - 1} = 0.2.\end{aligned}$$

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$$\frac{\partial^4 f}{\partial^4 x} = -\frac{6}{(x + 2y)^4}, \quad \frac{\partial^4 f}{\partial^4 y} = -\frac{96}{(x + 2y)^4}.$$

$$\left| \frac{\partial^4 f}{\partial^4 x} \right| \leq \frac{6}{(1.2 + 2 \times 0.2)^4} \approx 0.91553 \quad \text{for } (x, y) \in R,$$

$$\left| \frac{\partial^4 f}{\partial^4 y} \right| \leq \frac{96}{(1.2 + 2 \times 0.2)^4} \approx 14.648.$$

$$\begin{aligned}\text{Quad_Error} &= \frac{(b-a)(d-c)}{180} \left| k^4 \frac{\partial^4 f}{\partial^4 y}(\hat{\xi}, \hat{\eta}) + h^4 \frac{\partial^4 f}{\partial^4 x}(\xi, \eta) \right| \\ &\leq \frac{(2.4 - 1.2)(1 - 0.2)}{180} (0.2^4 \times 14.648 + 0.2^4 \times 0.91553)\end{aligned}$$

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Therefore

$$\left| \int \int_R \log(x + 2y) dA - \left(\sum_{i=1}^7 \sum_{j=1}^5 c_i \hat{c}_j f(x_i, y_j) \right) \right| = 1.546 \times 10^{-5}$$

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2-Dimensional Gaussian Quadratures

$$\begin{aligned} R &= \{(x, y) \mid a \leq x \leq b, c \leq y \leq d.\} \\ \int \int_R f(x, y) dA &= \int_a^b \left(\int_c^d f(x, y) dy \right) dx. \end{aligned}$$

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$$\int \int_R f(x, y) dA = \int_a^b \left(\int_c^d f(x, y) dy \right) dx.$$

Perform change of variables

$$x = \frac{a+b}{2} + \frac{b-a}{2} u, \quad y = \frac{c+d}{2} + \frac{d-c}{2} v \quad \text{for } u, v \in [-1, 1].$$

Double integral becomes

$$\int \int_R f(x, y) dA = \frac{(b-a)(d-c)}{4} \int_{-1}^1 \hat{g}(u) du, \quad \text{where}$$

$$\hat{g}(u) \stackrel{\text{def}}{=} \int_{-1}^1 f \left(\frac{a+b}{2} + \frac{b-a}{2} u, \frac{c+d}{2} + \frac{d-c}{2} v \right) dv.$$

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- n -point Gaussian quadrature for $\int_{-1}^1 \widehat{g}(u) du$:

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$$x_i = \frac{a+b}{2} + \frac{b-a}{2} u_i, \quad \widehat{g}(u_i) = \int_{-1}^1 f\left(x_i, \frac{c+d}{2} + \frac{d-c}{2} v\right) dv.$$

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- ▶ m -point Gaussian quadrature for $\widehat{g}(u_i)$,

$$\begin{aligned} \widehat{g}(u_i) &\approx \widehat{c}_1 f\left(x_i, \frac{c+d}{2} + \frac{d-c}{2} v_1\right) + \cdots + \widehat{c}_m f\left(x_i, \frac{c+d}{2} + \frac{d-c}{2} v_m\right) \\ &= \widehat{c}_1 f(x_i, y_1) + \cdots + \widehat{c}_m f(x_i, y_m), \quad y_j \stackrel{\text{def}}{=} \frac{c+d}{2} + \frac{d-c}{2} v_j. \end{aligned}$$

$$\int \int_R f(x, y) dA = \frac{(b-a)(d-c)}{4} \int_{-1}^1 \widehat{g}(u) du$$

$$\begin{aligned}\int_{-1}^1 \widehat{g}(u) du &\approx c_1 \widehat{g}(u_1) + c_2 \widehat{g}(u_2) + \cdots + c_n \widehat{g}(u_n) \\ \widehat{g}(u_i) &\approx \widehat{c}_1 f(x_i, y_1) + \cdots + \widehat{c}_m f(x_i, y_m).\end{aligned}$$

So we have Gaussian quadrature for double integral:

$$\int \int_R f(x, y) dA \approx \frac{(b-a)(d-c)}{4} \sum_{i=1}^n \sum_{j=1}^m c_i \widehat{c}_j f(x_i, y_j).$$

Example: $\int \int_R \log(x + 2y) dA = 1.036 \dots$, $n = 7, m = 5$

► Gaussian quadrature approximation

$$\int \int_R f(x, y) dA \approx \frac{(b-a)(d-c)}{4} \sum_{i=1}^n \sum_{j=1}^m c_i \hat{c}_j f(x_i, y_j) \approx \underline{1.03604817065 \dots}$$

$$\left| \int \int_R f(x, y) dA - \frac{(b-a)(d-c)}{4} \sum_{i=1}^n \sum_{j=1}^m c_i \hat{c}_j f(x_i, y_j) \right| \approx 6.4 \times 10^{-10}.$$

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► Simpson rule approximation

$$\int \int_R f(x, y) dA \approx \sum_{i=1}^n \sum_{j=1}^m c_i \hat{c}_j f(x_i, y_j) \approx \underline{1.03603270963 \dots}.$$

$$\left| \int \int_R f(x, y) dA - \sum_{i=1}^n \sum_{j=1}^m c_i \hat{c}_j f(x_i, y_j) \right| \approx 1.5 \times 10^{-5}.$$

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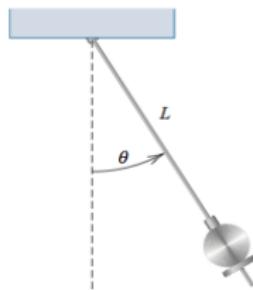
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Gaussian quadrature much more accurate.

§5.1 Initial Value ODE

- The motion of a swinging pendulum

$$\frac{d^2\theta}{dt^2} + \frac{g}{L} \sin \theta = 0,$$

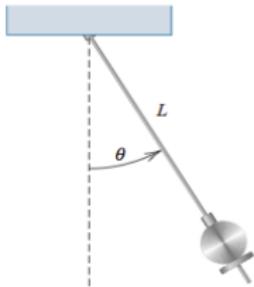


- Initial value conditions critical
 $\theta(t_0) = \theta_0$, and $\theta'(t_0) = \theta'_0$.

§5.1 Initial Value ODE

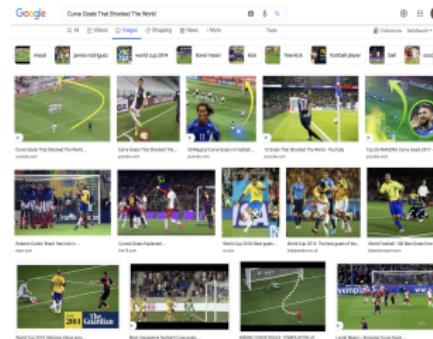
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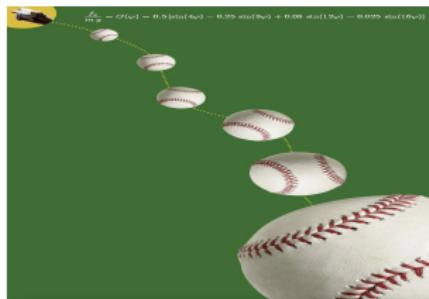


- ▶ Initial value conditions critical
 $\theta(t_0) = \theta_0$, and $\theta'(t_0) = \theta'_0$.

- #### ► Curveball/knuckleball in soccer



- #### ► Knuckleball pitch



Lipschitz condition

Definition: function $f(t, y)$ satisfies a **Lipschitz condition** in the variable y on a set $D \subset \mathbf{R}^2$ if a constant $L > 0$ exists with

$$|f(t, y_1) - f(t, y_2)| \leq L |y_1 - y_2|,$$

whenever $(t, y_1), (t, y_2)$ are in D . L is Lipschitz constant.

- ▶ **Example 1:** Show that $f(t, y) = t|y|$ satisfies a Lipschitz condition on the region

$$D = \{(t, y) \quad | \quad 0 \leq t \leq T\}.$$

Solution: For any $(t, y_1), (t, y_2)$ in D ,

$$|f(t, y_1) - f(t, y_2)| = |t|y_1| - t|y_2|| \leq t |y_1 - y_2| \leq L |y_1 - y_2|,$$

for $L = T$.

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- ▶ **Example 2:** Show that $f(t, y) = t y^2$ does not satisfy any Lipschitz condition on the region

$$D = \{(t, y) \mid 0 \leq t \leq T\}.$$

Solution: Choose $(T, y_1), (T, y_2)$ in D with $y_1 = 0, y_2 > 0$,

$$\frac{|f(T, y_1) - f(T, y_2)|}{|y_1 - y_2|} = T y_2,$$

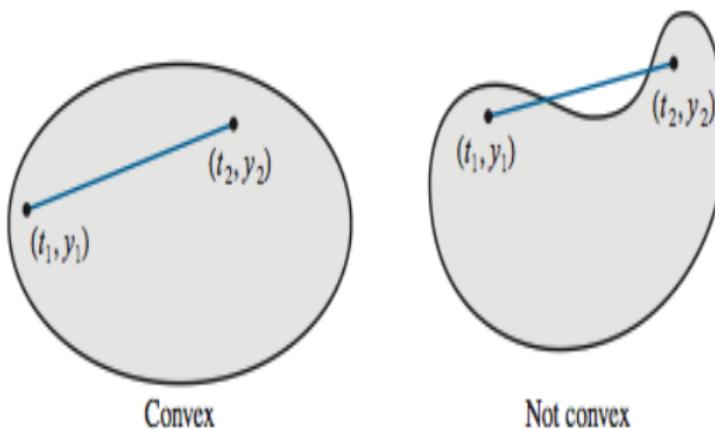
which can be larger than L for any fixed $L > 0$.

Convex Set

Definition: A set $D \subset \mathbf{R}^2$ is convex if

whenever (t_1, y_1) and $(t_2, y_2) \in D$

→ line segment $(1 - \lambda)(t_1, y_1) + \lambda(t_2, y_2) \in D$ for all $\lambda \in [0, 1]$.



Theorem: Suppose $f(t, y)$ is defined on a convex set $D \subset \mathbf{R}^2$. If a constant $L > 0$ exists with

$$\left| \frac{\partial f}{\partial y}(t, y) \right| \leq L, \quad \text{for all } (t, y) \in D,$$

then f satisfies a Lipschitz condition with Lipschitz constant L .

Theorem: Suppose $f(t, y)$ is defined on a convex set $D \subset \mathbf{R}^2$. If a constant $L > 0$ exists with

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then f satisfies a Lipschitz condition with Lipschitz constant L .

- ▶ **Example 1:** Show that $f(t, y) = t y^2$ satisfies Lipschitz condition on the region

$$D = \{(t, y) \mid 0 \leq t \leq T, -Y \leq y \leq Y\}.$$

Solution:

$$\frac{\partial f}{\partial y}(t, y) = 2ty, \quad \left| \frac{\partial f}{\partial y}(t, y) \right| \leq 2T|y| \quad \text{for all } (t, y) \in D.$$

so $f(t, y) = t y^2$ satisfies Lipschitz condition with $L = 2T|Y|$.

What is special with $f(t, y) = t y^2$?

- ▶ $f(t, y) = t y^2$ satisfies Lipschitz condition on the region

$$D = \{(t, y) \mid 0 \leq t \leq T, -Y \leq y \leq Y\}.$$

- ▶ $f(t, y) = t y^2$ doesn't satisfy Lipschitz condition on region

$$D = \{(t, y) \mid 0 \leq t \leq T\}.$$

Initial value problem

$$y'(t) = t y^2(t), \quad y(t_0) = \alpha > 0$$

has unique, but unbounded solution

$$y(t) = \frac{2\alpha}{2 + \alpha(t_0^2 - t^2)},$$

the denominator vanishes at

$$t = \sqrt{\frac{2}{\alpha} + t_0^2}.$$

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- ▶ for $0 < t_0 < T < \sqrt{\frac{2}{\alpha} + t_0^2}$, ODE solution bounded on

$$D = \{(t, y) \mid t_0 \leq t \leq T\}.$$

- ▶ for $\sqrt{\frac{2}{\alpha} + t_0^2} < T$ ODE solution not defined at $t = \sqrt{\frac{2}{\alpha} + t_0^2}$.

Well-posed problem

Definition in English: ODE is well-posed if

- ▶ A unique ODE solution exists, and
- ▶ Small changes (perturbation) to ODE imply small changes to solution.

Well-posed problem

The initial-value problem

$$\frac{dy}{dt} = f(t, y), \quad a \leq t \leq b, \quad y(a) = \alpha, \quad (5.2)$$

is said to be a **well-posed problem** if:

- A unique solution, $y(t)$, to the problem exists, and
- There exist constants $\varepsilon_0 > 0$ and $k > 0$ such that for any ε , with $\varepsilon_0 > \varepsilon > 0$, whenever $\delta(t)$ is continuous with $|\delta(t)| < \varepsilon$ for all t in $[a, b]$, and when $|\delta_0| < \varepsilon$, the initial-value problem

$$\frac{dz}{dt} = f(t, z) + \delta(t), \quad a \leq t \leq b, \quad z(a) = \alpha + \delta_0, \quad (5.3)$$

has a unique solution $z(t)$ that satisfies

$$|z(t) - y(t)| < k\varepsilon \quad \text{for all } t \text{ in } [a, b].$$



Well-posed problem

Definition in English: ODE is well-posed if

- ▶ A unique ODE solution exists, and
- ▶ Small changes (perturbation) to ODE imply small changes to solution.

Theorem: Suppose

$$\mathcal{D} = \{(t, y) \mid a \leq t \leq b \text{ and } -\infty < y < \infty\}.$$

If f is continuous and satisfies a Lipschitz condition in the variable y on the set \mathcal{D} , then the initial-value problem

$$\frac{dy}{dt} = f(t, y), \quad a \leq t \leq b, \quad y(a) = \alpha$$

is well-posed.

Well-posed problem, example

Show that the initial-value problem

$$\frac{dy}{dx} = y - t^2 + 1, \quad 0 \leq t \leq 2,$$

is well posed on

$$\mathcal{D} \stackrel{\text{def}}{=} \{(t, y) \mid 0 \leq t \leq 2 \text{ and } -\infty < y < \infty\}$$

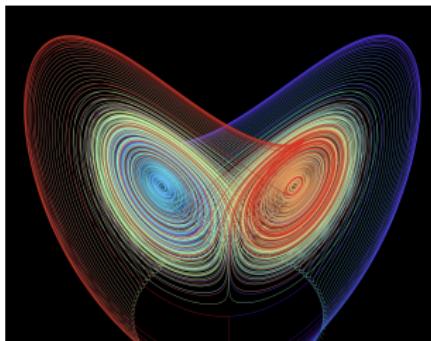
Solution: Because

$$\frac{\partial f}{\partial y}(t, y) = 1, \quad \left| \frac{\partial f}{\partial y}(t, y) \right| = 1.$$

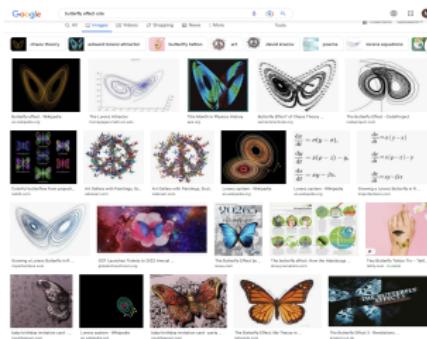
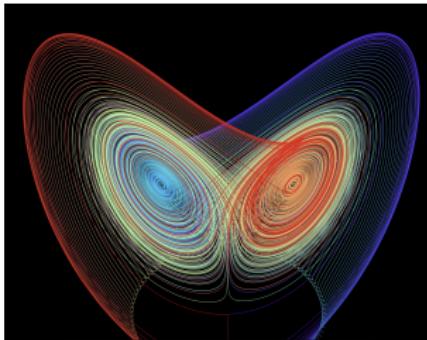
$f(t, y) = y - t^2 + 1$ satisfies a Lipschitz condition in y on \mathcal{D} with Lipschitz constant 1. Therefore this ODE is well-posed. In fact,

$$y(t) = 1 + t^2 + 2t - \frac{1}{2}e^t.$$

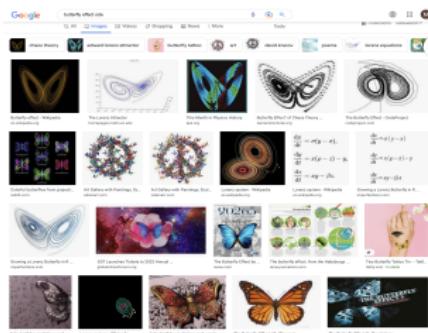
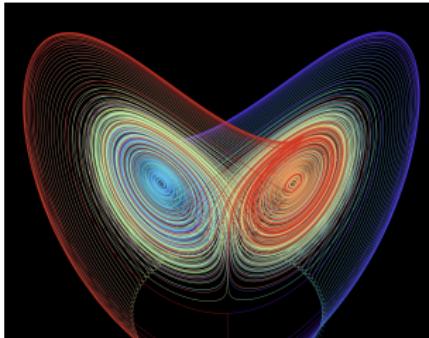
What is **not** well-posed?



What is **not** well-posed?



What is **not** well-posed?



en.wikipedia.org/wiki/Lorenz_system#MATLAB_simulation

MATLAB simulation [edit]

```
# Define over time interval [0,100] with initial conditions (L,L,I)
# 't' is use of differential equation
# 'x' is array containing x, y, z variables
# 't' is time variable

sigma = 10;
beta = 8/3;
rho = 28;

dx = sigma*(x(1)-x(2));
dy = (x(1)*x(3)-x(1)*x(2))-beta*x(2);
dz = (x(1)*x(2)-rho*x(3));
[t,x] = odeset(t,[0 100],[1 1 1]); % Runge-Kutta 4th/5th order ODE solver
plot3(x(1),x(2),x(3),t)
```

Mathematica simulation [edit]

Standard way:

```
#dx = x[1];
#y = x[2];
T[1] := x[1](t - \[Delta]t);
T[1] := x[1](t - \[Delta]t) - x[1]t;
x[1] := T[1];
x[2] := 10;
x[3] := 28;
x[1] := 10; y[1] := 10; z[1] := 10;
pars = {x[1], y[1], z[1], \[Delta]t};
{x, y, z} =
Module[{x[1], y[1], z[1], \[Delta]t}, {x[1], y[1], z[1]}];
ParametricPlot3D[{x[1], x[2], x[3]}, {t, 0, 100}]
```

Less verbose:

```
lorenz = Module[state0, Print[Module[{{x = 10, y = 10, z = 10, \[Delta]t = 0.01}}, state0 = StateSpace[Lorenz, {{x, 10, 10}, {y, 10, 10}, {z, 10, 10}}]; ParametricPlot3D[Lorenz, {t, 0, 100}]]];
```

Python simulation [edit]

```
# Import numpy as np
import numpy as np
from scipy.integrate import odeint
from scipy.integrate import ode

rho = 28.0
sigma = 10.0
beta = 8.3

def lorenz(t, u):
    x, y, z = u
    # Derivatives of the state vector
    return sigma * (y - x), x * (rho - z) - y, x * y - beta * z

state0 = [10.0, 10.0, 10.0]
t = np.arange(0.0, 40.0, 0.01)

state = odeint(lorenz, state0, t)

fig = plt.figure()
ax = fig.gca(projection='3d')
ax.plot(state[:, 0], state[:, 1], state[:, 2])
plt.show()
```