
NASDE 25.10.2024



Today: \Rightarrow Proof of Marsaglia Polar.

\rightarrow General normal d-n

\rightarrow Integration methods

Proof of Marsaglia Polar:

Take any measurable $g: \mathbb{R}^2 \rightarrow \mathbb{R}$

$$\mathbb{E}[g(X)] = \mathbb{E}[g \circ f(U)]$$

$$= \frac{1}{\pi \cdot \mathbb{1}_{B(0,1)}} \iint_{B(0,1)} g\left(\frac{u_1}{\sqrt{u_1^2 + u_2^2}} \cdot \sqrt{-2 \ln(u_1^2 + u_2^2)}, \frac{u_2}{\sqrt{u_1^2 + u_2^2}} \cdot \sqrt{-2 \ln(u_1^2 + u_2^2)}\right) du_1 du_2$$

Change to Polar coordinates: $(u_1, u_2) = (r \cos \varphi, r \sin \varphi)$, $u_1^2 + u_2^2 = r^2$
 $du_1 du_2 = r dr d\varphi$

$$= \frac{1}{\pi} \int_0^{2\pi} \int_0^{\infty} g\left(\frac{r \cos \varphi}{r} \cdot \sqrt{-2 \ln(r^2)}, \frac{r \sin \varphi}{r} \cdot \sqrt{-2 \ln(r^2)}\right) r dr d\varphi$$

$$= \frac{1}{\pi} \int_0^{2\pi} \int_0^{\infty} g(\cos \varphi \cdot 2 \cdot \sqrt{-\ln(r)}, \sin \varphi \cdot 2 \cdot \sqrt{-\ln(r)}) r dr d\varphi$$

Change of variables: $s = -\ln(r) \Rightarrow r = e^{-s}, dr = -e^{-s} ds \Rightarrow$

$$= \frac{1}{\pi} \int_0^{2\pi} \int_0^{\infty} g(\cos \varphi \cdot 2 \cdot \sqrt{s}, \sin \varphi \cdot 2 \cdot \sqrt{s}) e^{-2s} ds d\varphi$$

COV: $2\sqrt{s} = \tilde{r} \rightarrow \frac{d\tilde{r}}{ds} = \frac{1}{\sqrt{s}} \frac{ds}{\frac{\tilde{r}}{2}} d\tilde{r}$
 $ds = \sqrt{s} d\tilde{r} = \frac{\tilde{r}}{2} d\tilde{r}$

$$= \frac{1}{\pi} \int_0^{2\pi} \int_0^{\infty} g(\cos \varphi \cdot \tilde{r}, \sin \varphi \cdot \tilde{r}) e^{-\frac{1}{2} \tilde{r}^2} \cdot \frac{\tilde{r}}{2} d\tilde{r} d\varphi$$

$(\tilde{r} \cos \varphi, \tilde{r} \sin \varphi) \rightarrow (x_1, x_2)$
 $\tilde{r} d\tilde{r} d\varphi = dx_1 dx_2$

$$= \frac{1}{2\pi} \iint_{\mathbb{R}^2} g(x_1, x_2) \cdot e^{-\frac{1}{2}(x_1^2 + x_2^2)} \cdot dx_1 dx_2 = \mathbb{E}[g(Y_1, Y_2)], (Y_1, Y_2) \sim \mathcal{N}(0, I_{\mathbb{R}^2})$$

$$\mathbb{E}[g(x)] = \mathbb{E}[g(Y_1, Y_2)] \quad (3)$$

$$(Y_1, Y_2) \sim \mathcal{N}(0, \mathbb{I}_{\mathbb{R}^2})$$

$\forall g$ -measurable. \Rightarrow

$$X = (X_1, X_2) \sim \mathcal{N}(0, \mathbb{I}_{\mathbb{R}^2}) \quad \square$$

Generate normal d-n with general mean and variance:

If $X \sim \mathcal{N}(0, \mathbb{I}_{\mathbb{R}^d})$, $b \in \mathbb{R}^d$,
 $A \in \mathbb{R}^{d \times d}$

$$AX + b \sim \mathcal{N}_{b, AA^T}$$

Assume: $b \in \mathbb{R}^d$, $g \in \mathbb{R}^{d \times d}$ strictly positive symmetric

Write

$$g = L L^T \text{ - Cholesky decompos. } (4)$$

„chol“ in Matlab.

Set $X = L \cdot X + b \Rightarrow$

$$X \sim \mathcal{N}(b, L L^T) = \mathcal{N}(b, g)$$

Integration methods:

In the context of SDEs we are interested in approximations

$$X^n: [0, T] \times \Omega \rightarrow \mathbb{R} \text{ for}$$

$X: [0, T] \times \Omega \rightarrow \mathbb{R}$ sol-n process of SDE.

• pathwise approximation:
Fix $\omega \in \Omega$.

(5)

$X^n(\omega): [0, T] \rightarrow \mathbb{R}$ to be close
to $X(\omega): [0, T] \rightarrow \mathbb{R}$

• approximation of statistical values
 $E[X]$, $\text{var}(X)$, ...

\Rightarrow numerical integration
(approximations of integrals)

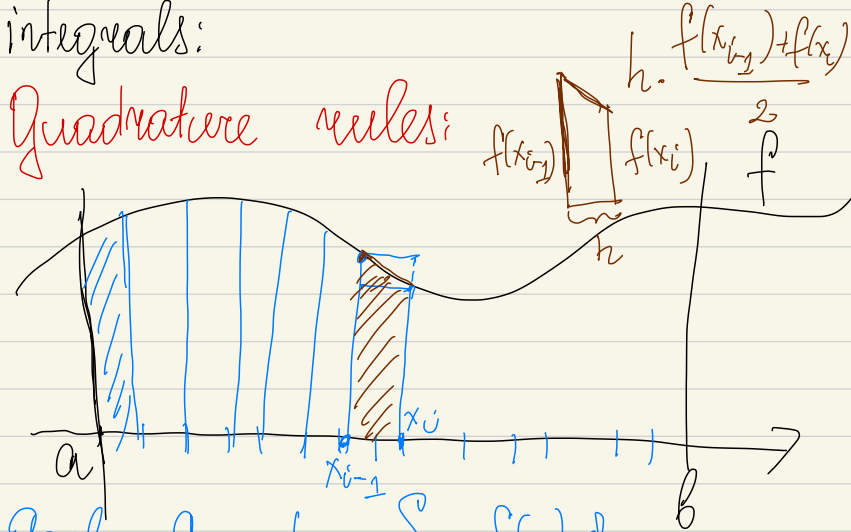
\rightarrow deterministic numerical
integration

(usually suffer from the curse
of dimensionality)

\rightarrow Monte Carlo methods (6)
 \rightarrow usually better alternative in high
dim- Ω s

\rightarrow can also be used for deterministic
integrals:

Quadrature rules:



Goal: Compute $\int_{[a, b]} f(x) dx$

A quadrature rule Q on $A \subseteq \mathbb{R}^d$
is given by
• $X = \{x_i, i \in I\}$, $\#I < \infty$ nodes

weights $w = \{w_i, i \in \mathbb{I}\}$ (7)

We approximate $\int_A f(x) dx$ by

$$Q[f] := \sum_{i \in \mathbb{I}} w_i \cdot f(x_i) \approx \int_A f(x) dx$$

a) Left rectangle method
Equidistant nodes $h = \frac{b-a}{n}, n \in \mathbb{N}$

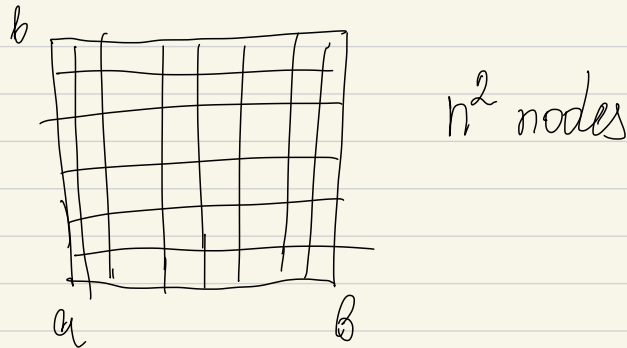
Nodes: $\mathcal{X} = \{a, a+h, a+2h, \dots, a+(n-1) \cdot h\}$
 $= \{a+i \cdot h, i=0, \dots, n-1\}$

[Right rectangle: $\mathcal{X} = \{a+h, a+2h, \dots, a+n \cdot h = b\}$]

Equal weights $w_1 = w_2 = \dots = w_n = h \Rightarrow$

$$R_{[a,b]}^n [f] = h \sum_{i=0}^{n-1} f(a+i \cdot h)$$

In 2d: $[a,b] \times [a,b]$ (8)



Note: in case of d-dimension:

$$\mathcal{X} = \{ \underbrace{(a+i_1 \cdot h, a+i_2 \cdot h, \dots, a+i_d \cdot h)}_{i_1, i_2, \dots, i_d \in \{0, 1, \dots, n-1\}} \}$$

Weights:

$$w_{i_1, i_2, \dots, i_d} = h^d, i_1, i_2, \dots, i_d \in \{0, \dots, n-1\}$$

$$R_{[a,b]^d}^n = h^d \left(\sum_{i_1=0}^{n-1} \dots \sum_{i_d=0}^{n-1} f(x_{i_1, i_2, \dots, i_d}) \right)$$

⑧ Trapezoidal rule:

$$P^n_{[a,b]}[f] := \sum_{i=0}^{n-1} h \cdot \frac{f(a+ih) + f(a+(i+1)h)}{2}$$

$$= h \cdot \left(\sum_{i=0}^{n-1} \frac{f(a+ih) + f(a+(i+1)h)}{2} \right)$$

$$= h \left(\frac{f(a) + f(a+h)}{2} + \frac{f(a+h) + f(a+2h)}{2} + \dots \right)$$

$$+ \frac{f(a+(n-1)h) + f(a+nh)}{2}$$

$$= h \cdot \left(\frac{f(a) + f(b)}{2} + \sum_{i=1}^{n-1} f(a+ih) \right)$$

$$h = \frac{b-a}{n}$$

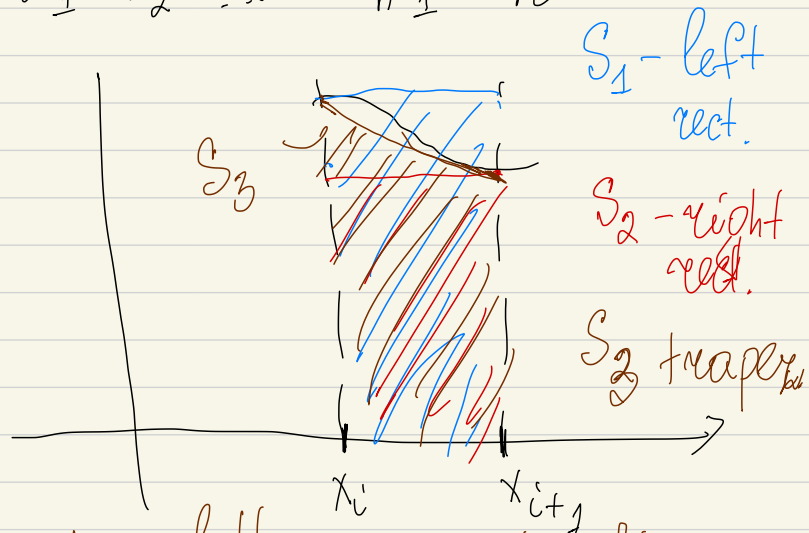
⑨

$$x = \{a+ih, i=0, \dots, n\}$$

$$w_0 = w_n = \frac{h}{2}$$

$$w_1 = w_2 = \dots = w_{n-1} = h$$

⑩



S_3 is a better approximation than S_1 and S_2 .

To analyze the errors we need (11) notions of modulus of continuity and Hölder continuity

Def-n: Let (E, d_E) and (F, d_F) be metric spaces and $f: E \rightarrow F$

$w_f: [0, \infty] \rightarrow [0, \infty]$ is modulus of continuity of f given by

$$w_f(h) = \sup_{\substack{x, y \in E \\ d_E(x, y) \leq h}} d_F(f(x), f(y))$$

Remark: f is uniformly continuous

$$\Leftrightarrow w_f(h) \xrightarrow{h \rightarrow 0} 0$$

will be in Ex sheet 3.

Def-n: Let $\alpha \in (0, 1]$
 $\|f\|_{C^\alpha(E, F)} = \sup_{\substack{x, y \in E \\ x \neq y}} \frac{d_F(f(x), f(y))}{|d_E(x, y)|^\alpha}$ (12)

$C^\alpha(E, F) = \{f: E \rightarrow F, \text{ such that}$

$$\|f\|_{C^\alpha(E, F)} < \infty\}$$

Space of Hölder cont. f-ns

Remark:

$$\|f\|_{C^\alpha} = \sup_{h \in (0, \infty)} \left[\frac{w_f(h)}{h^\alpha} \right]$$

$\alpha = 1$: Lipschitz continuity.
Next time: Analyse error of $R^n[f]$