

# *Numerical Integration*

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# Simple Monte Carlo Integration (NRiC §7.6)

- Can use RNGs to estimate integrals.
- Suppose we pick  $n$  random points  $\mathbf{x}_1, \dots, \mathbf{x}_N$  uniformly in a multi-D volume  $V$ .
- Basic theorem of Monte Carlo integration:

$$\int_V f dV \simeq V \langle f \rangle \pm V \sqrt{\frac{\langle f^2 \rangle - \langle f \rangle^2}{N}},$$

where

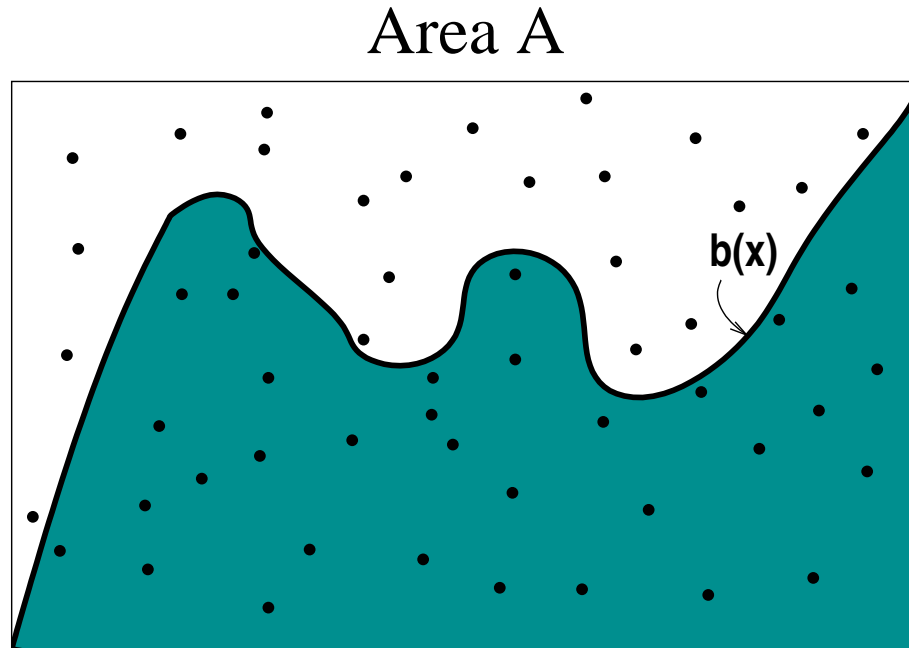
$$\langle f \rangle \equiv \frac{1}{N} \sum_{i=1}^N f(\mathbf{x}_i) \quad \text{and} \quad \langle f^2 \rangle \equiv \frac{1}{N} \sum_{i=1}^N f^2(\mathbf{x}_i).$$

- The  $\pm$  term is a  $1\text{-}\sigma$  error estimate, not a rigorous bound.

- Previous formula works fine if  $V$  is simple.
- What if we want to integrate a function  $g$  over a region  $W$  that is *not* easy to sample randomly?
- Solution: find a simple volume  $V$  that *encloses*  $W$  and define a new function  $f(\mathbf{x})$ ,  $\mathbf{x} \in V$ , such that:

$$f(\mathbf{x}) = \begin{cases} g(\mathbf{x}) & \text{for all } \mathbf{x} \in W, \\ 0 & \text{otherwise.} \end{cases}$$

E.g., suppose we want to integrate  $g(x, y)$  over the shaded area inside area  $A$  below:



To integrate, take random samples over the whole rectangle, set

$$f(x_i, y_i) = \begin{cases} g(x_i, y_i) & y_i \leq b(x_i), \\ 0 & \text{otherwise,} \end{cases}$$

... and compute

$$\int_{\text{shaded area}} g(x, y) dx dy \simeq \frac{A}{N} \sum_i f(x_i, y_i).$$

- Nifty example:  $\pi$  can be estimated by integrating

$$p(x, y) = \begin{cases} 1 & x^2 + y^2 \leq 1, \\ 0 & \text{otherwise,} \end{cases}$$

over a  $2 \times 2$  square:

$$\begin{aligned} \pi &= \int_{-1}^1 \int_{-1}^1 p(x, y) dx dy \\ &\simeq \frac{4}{N} \sum_i p(x_i, y_i). \end{aligned}$$

- See *NRiC* for another worked example.

- Optimization strategy: make  $V$  as close as possible to  $W$ , since zero values of  $f$  will increase the *relative* error estimate.
- Principal disadvantage: accuracy increases only as square root of  $N$ .
- Fancier routines exist for faster convergence: *NRiC* §7.7–7.8.
- Monte Carlo techniques used in a variety of other contexts: anywhere statistical sampling is useful. E.g.,
  - Monte Carlo chains to find best fit parameters in multi-D space
  - Simulating scattering processes (E.g. photon scattering)
  - Determining model fit significance by testing the model against many sets of random synthetic data with the same mean and variance.

# Numerical Integration (Quadrature)

- NRiC §4.
- Already seen Monte Carlo integration.
- Can cast problem as a differential equation (DE):

$$I = \int_a^b f(x) dx$$

is equivalent to solving for  $I \equiv y(b)$  the DE  $dy/dx = f(x)$  with the boundary condition (BC)  $y(a) = 0$ .

- Will learn about ODE solution methods next class.

# Trapezoidal and Simpson's rules

- Have abscissas  $x_i = x_0 + ih$ ,  $i = 0, 1, \dots, N + 1$ .
- A function  $f(x)$  has known values  $f(x_i) = f_i$ .
- Want to integrate  $f(x)$  between endpoints  $a$  and  $b$ .
- Trapezoidal rule (2-point closed formula):

$$\int_{x_1}^{x_2} f(x) dx = h \left[ \frac{1}{2} f_1 + \frac{1}{2} f_2 \right] + \mathcal{O}(h^3 f''),$$

i.e., the area of a trapezoid of base  $h$  and vertex heights  $f_1$  and  $f_2$ .

- Simpson's rule (3-point closed formula):

$$\int_{x_1}^{x_3} f(x) dx = h \left[ \frac{1}{3} f_1 + \frac{4}{3} f_2 + \frac{1}{3} f_3 \right] + \mathcal{O}(h^5 f^{(4)}).$$



```

#define FUNC(x) ((*func)(x))

float trapzd(float (*func)(float), float a, float b, int n)
{
float x,tnm,sum,del;
static float s;
int it,j;

if (n == 1) {
return (s=0.5*(b-a)*(FUNC(a)+FUNC(b)));
} else {
for (it=1,j=1;j<n-1;j++) it <<= 1;
tnm=it;
del=(b-a)/tnm;
x=a+0.5*del;
for (sum=0.0,j=1;j<=it;j++,x+=del) sum += FUNC(x);
s=0.5*(s+(b-a)*sum/tnm);
return s;
}
}
#undef FUNC
/* (C) Copr. 1986-92 Numerical Recipes Software ?421.1-9. */

```

# Extended trapezoidal rule

- If we apply the trapezoidal rule  $N - 1$  times and add the results, we get:

$$\int_{x_1}^{x_N} f(x) dx = h \left[ \frac{1}{2} f_1 + f_2 + f_3 + \dots + f_{N-1} + \frac{1}{2} f_N \right] + \mathcal{O} \left[ \frac{(b - a)^3 f''}{N^2} \right].$$

- Big advantage is it builds on previous work:
  - Coarsest step: average  $f$  at endpoints  $a$  and  $b$ .
  - Next refinement: add value at midpoint to average.
  - Next: add values at  $\frac{1}{4}$  and  $\frac{3}{4}$  points.
  - And so on. This is implemented as `trapzd()` in *NRiC*.

# More sophistication

- Usually don't know  $N$  in advance, so iterate to a desired accuracy: `qtrap()`.
- Higher-order method by cleverly adding refinements to cancel error terms: `qsimp()`.
- Generalization to order  $2k$  (*Richardson's deferred approach to the limit*): `qromb()`.
  - Uses extrapolation methods to set  $h \rightarrow 0$ .
- For improper integrals, generally need *open formulae* (not evaluated at endpoints).
- For multi- $D$ , use nested 1- $D$  techniques.

```

#include <math.h>
#define EPS 1.0e-6
#define JMAX 20
#define JMAXP (JMAX+1)
#define K 5

float qromb(float (*func)(float), float a, float b)
{
void polint(float xa[], float ya[], int n, float x, float *y, float *dy);
float trapzd(float (*func)(float), float a, float b, int n);
void nrerror(char error_text[]);
float ss,dss;
float s[JMAXP+1],h[JMAXP+1];
int j;

h[1]=1.0;
for (j=1;j<=JMAX;j++) {
s[j]=trapzd(func,a,b,j);
if (j >= K) {
polint(&h[j-K],&s[j-K],K,0.0,&ss,&dss);
if (fabs(dss) < EPS*fabs(ss)) return ss;
}
s[j+1]=s[j];
h[j+1]=0.25*h[j];
}
nrerror("Too many steps in routine qromb");
return 0.0;
}
#undef EPS
#undef JMAX
#undef JMAXP
#undef K
/* (C) Copr. 1986-92 Numerical Recipes Software ?421.1-9. */

```