

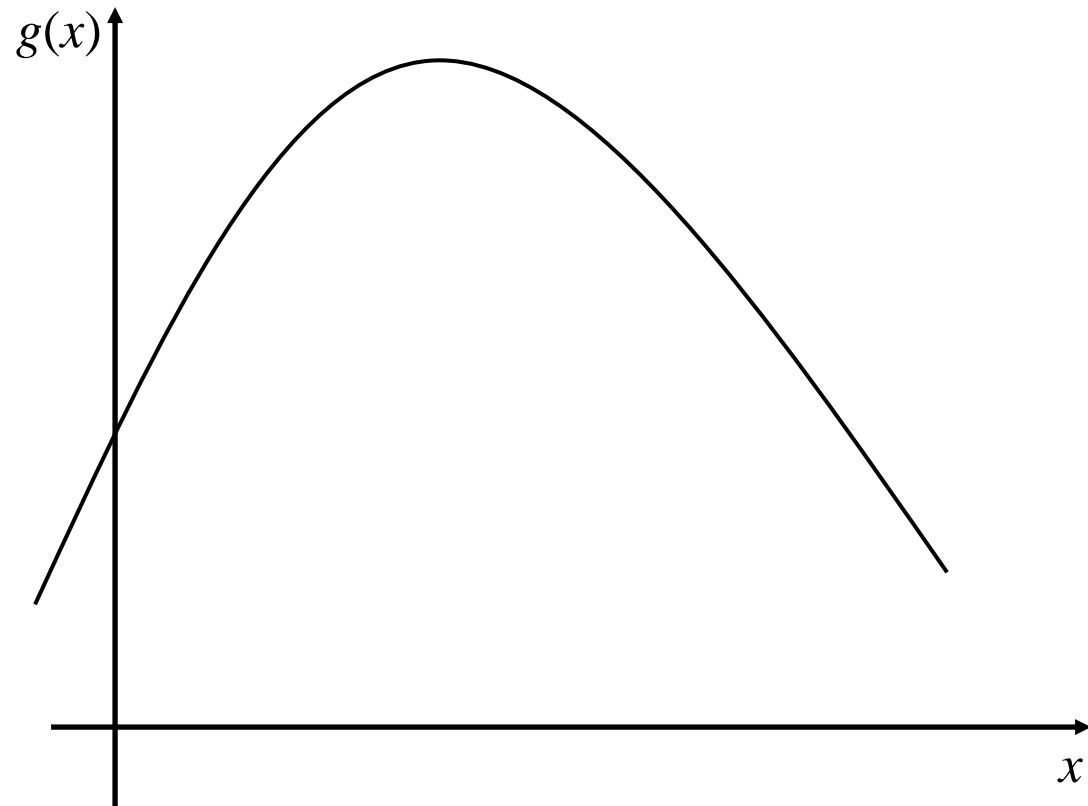
Numerical [^]Integration

Deterministic

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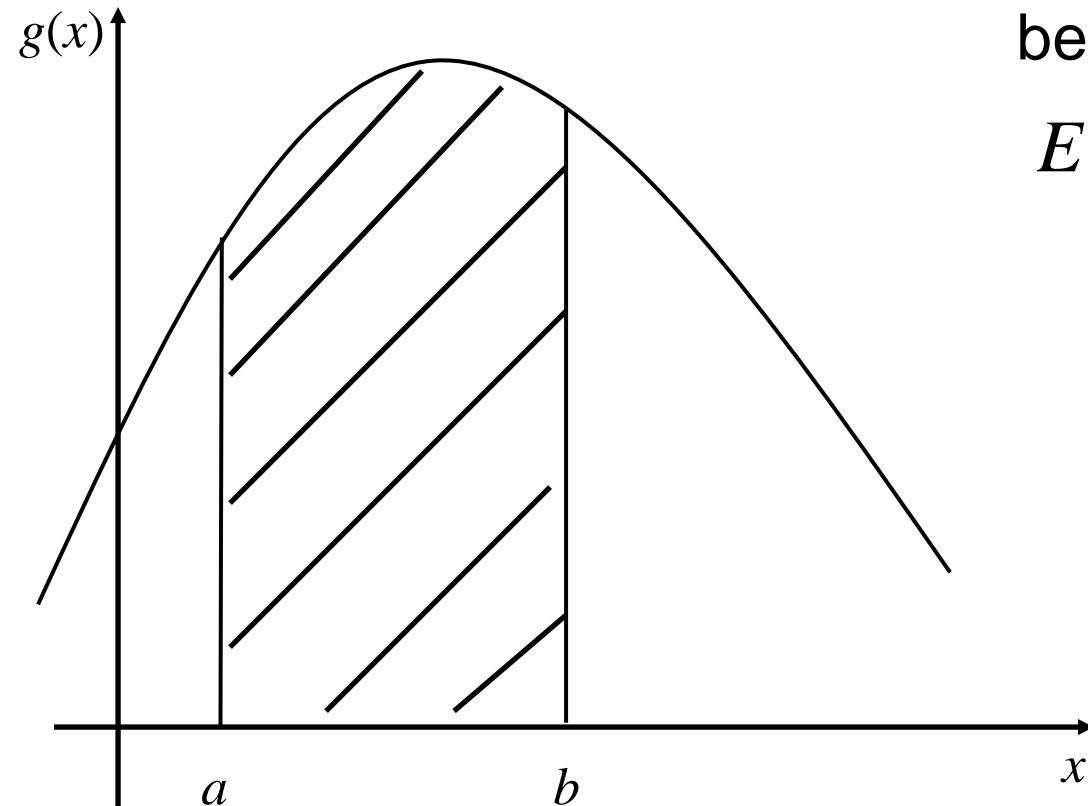
Deterministic Integration



goal

$$E(x) = \int_x g(x) dx$$

Deterministic Integration



Area under curve
between a and b .

$$E(x) = \int_x g(x) dx$$

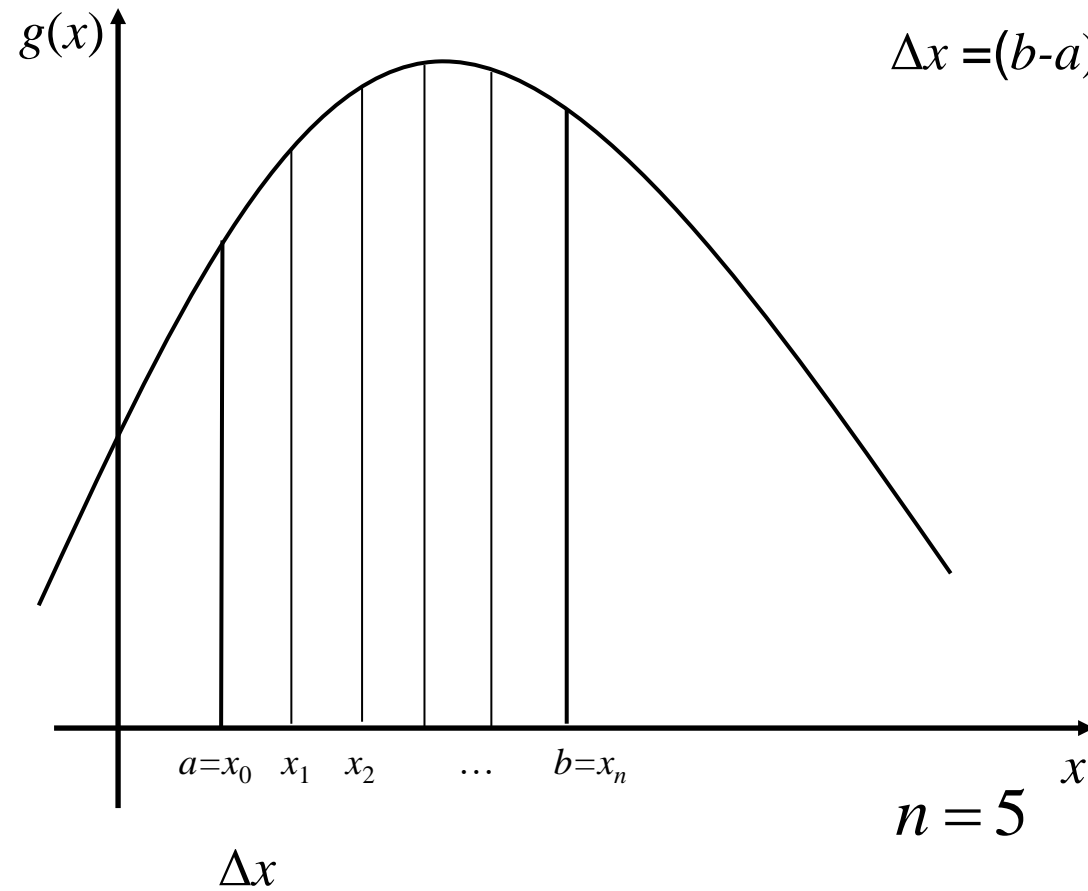
goal

$$E(x) = \int_x g(x) dx$$

think of as

$$g(x) = xf(x)$$

Deterministic Integration



Divide into intervals: Δx small

$$\Delta x = (b-a)/n \quad \Delta x = x_i - x_{i-1}$$

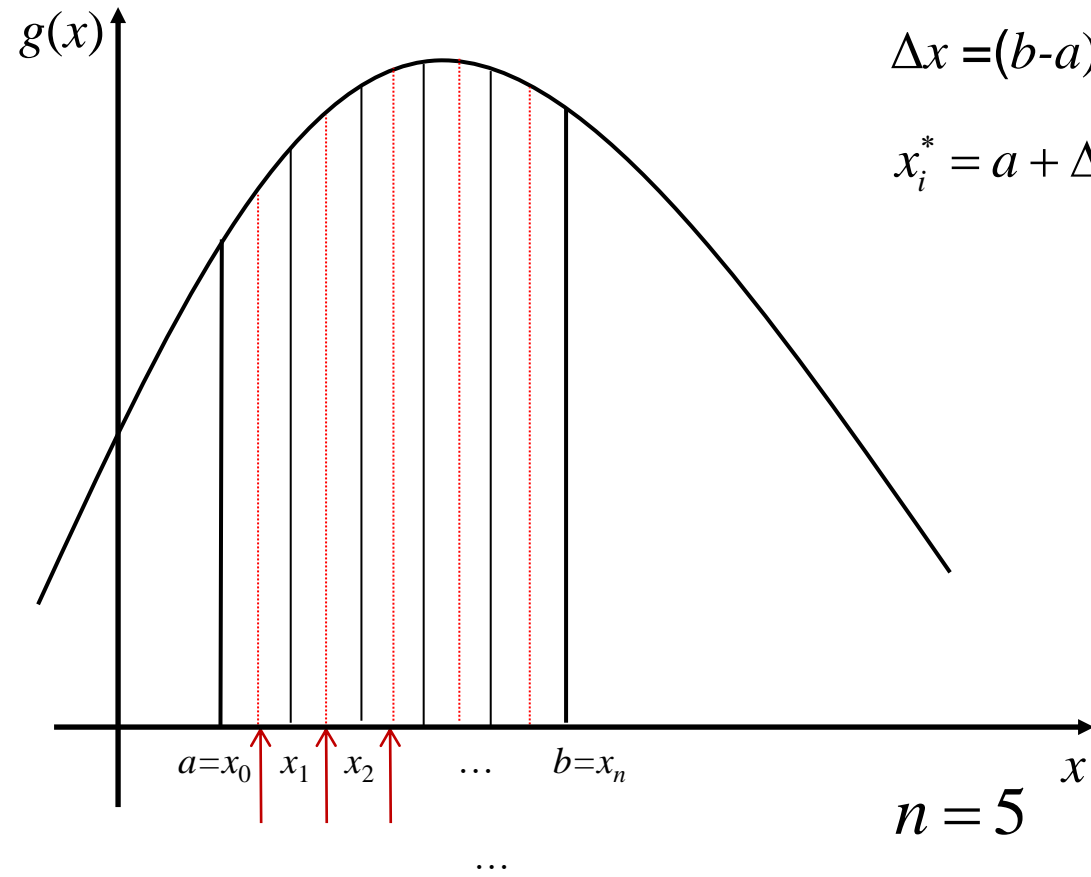
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think of as

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Deterministic Integration



Divide into intervals: Δx small

$$\Delta x = (b-a)/n \quad \Delta x = x_i - x_{i-1}$$

$$x_i^* = a + \Delta x / 2 + (i-1)\Delta x$$

$$i = 1, \dots, n$$

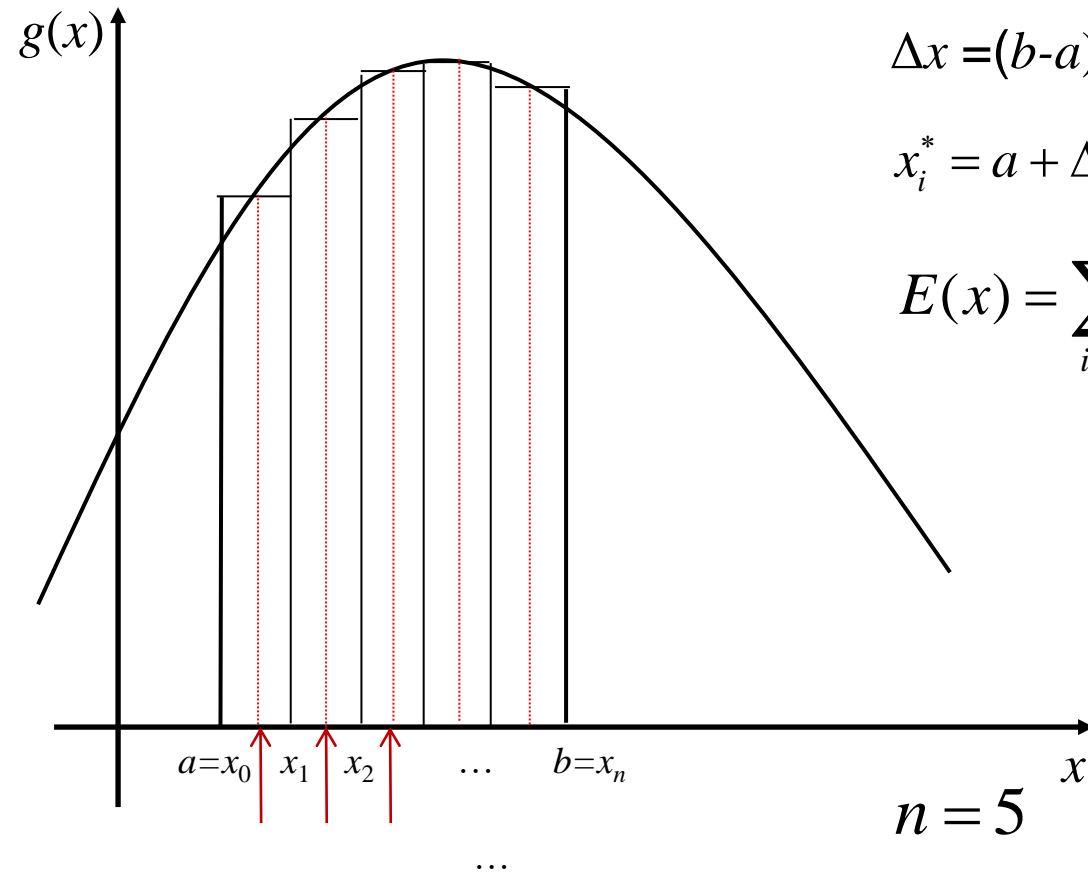
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$$E(x) = \int_x g(x)dx$$

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Deterministic Integration



Divide into intervals: Δx small

$$\Delta x = (b-a)/n \quad \Delta x = x_i - x_{i-1}$$

$$x_i^* = a + \Delta x / 2 + (i-1)\Delta x$$

$$E(x) = \sum_{i=1}^n g(x_i^*)\Delta x \quad i = 1, \dots, n$$

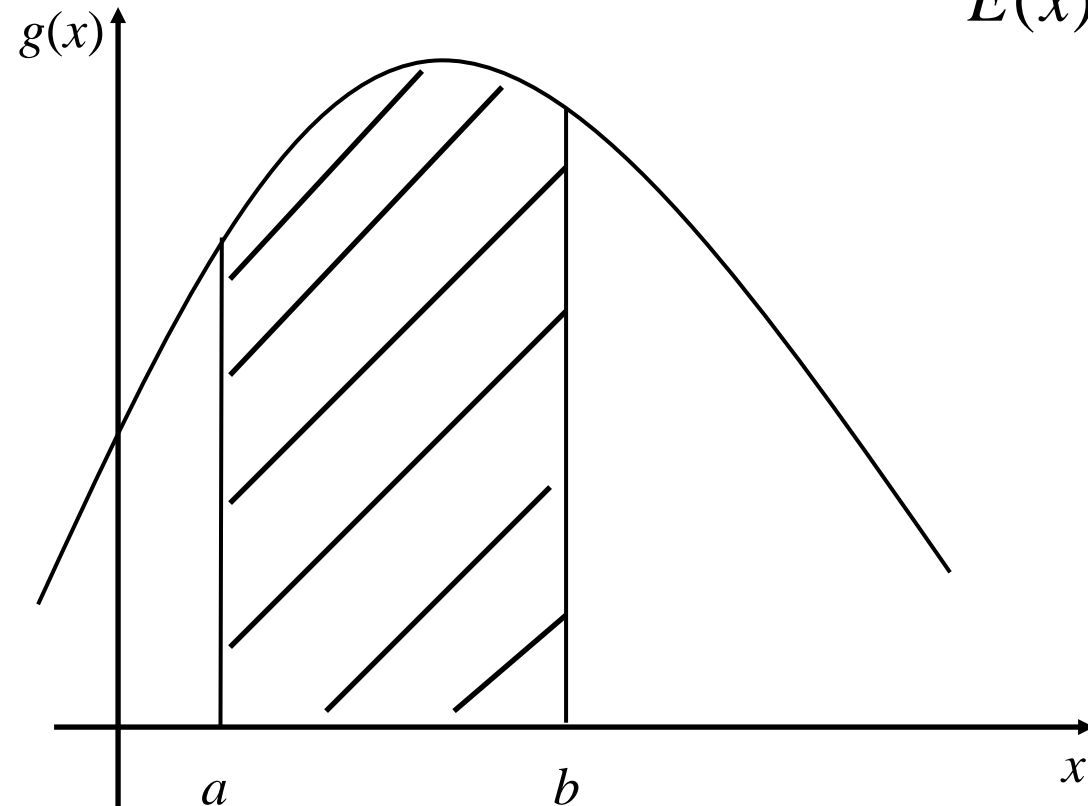
goal

$$E(x) = \int_x g(x)dx$$

think of as

$$g(x) = xf(x)$$

Deterministic Integration



$$E(x) = \lim_{\Delta x \rightarrow 0} \sum_{i=1}^n g(x_i^*) \Delta x$$

$$= \int_{x=a}^b g(x) dx$$

$$\Delta x = (b - a) / n$$

$$x_i^* = a + \Delta x / 2 + (i - 1)\Delta x$$

goal

$$E(x) = \int_x g(x) dx$$

think of as

$$g(x) = xf(x)$$

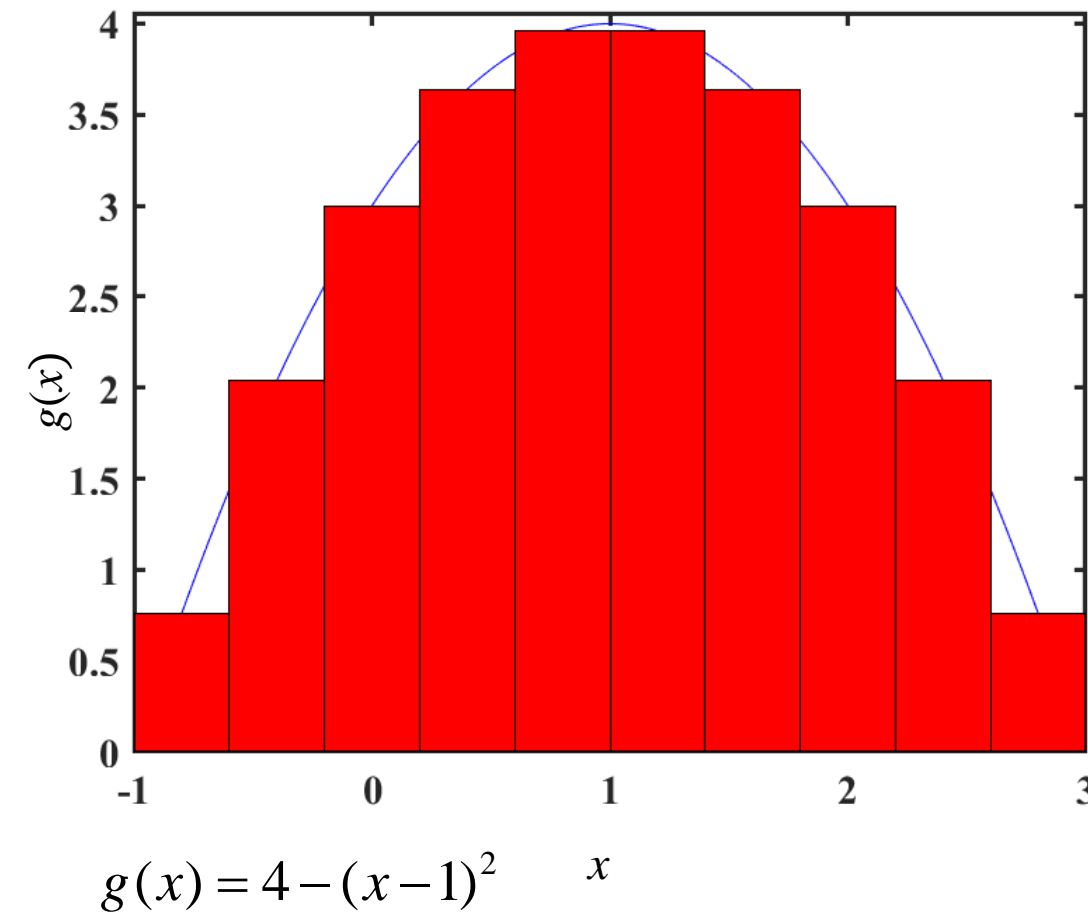
Deterministic Integration

goal

$$E(x) = \int_x g(x)dx$$

think of as

$$g(x) = xf(x)$$



← numerical

$$n=10, \quad \Delta x=0.400$$

$$\hat{E}(x) = \Delta x \sum_{i=1}^{10} g(x_i^*) = 10.7200$$

analytic

$$\int_{x=-1}^3 [4 - (x-1)^2] dx = 10.6667$$

```
% Numerical Integral
a=-1; b=3;
n=10; dx=(b-a)/n;
tpts=(a+dx/2:dx:b)';
gpts=4-(xpts-1).^2;
Exhat=dt*sum(gpts)
```

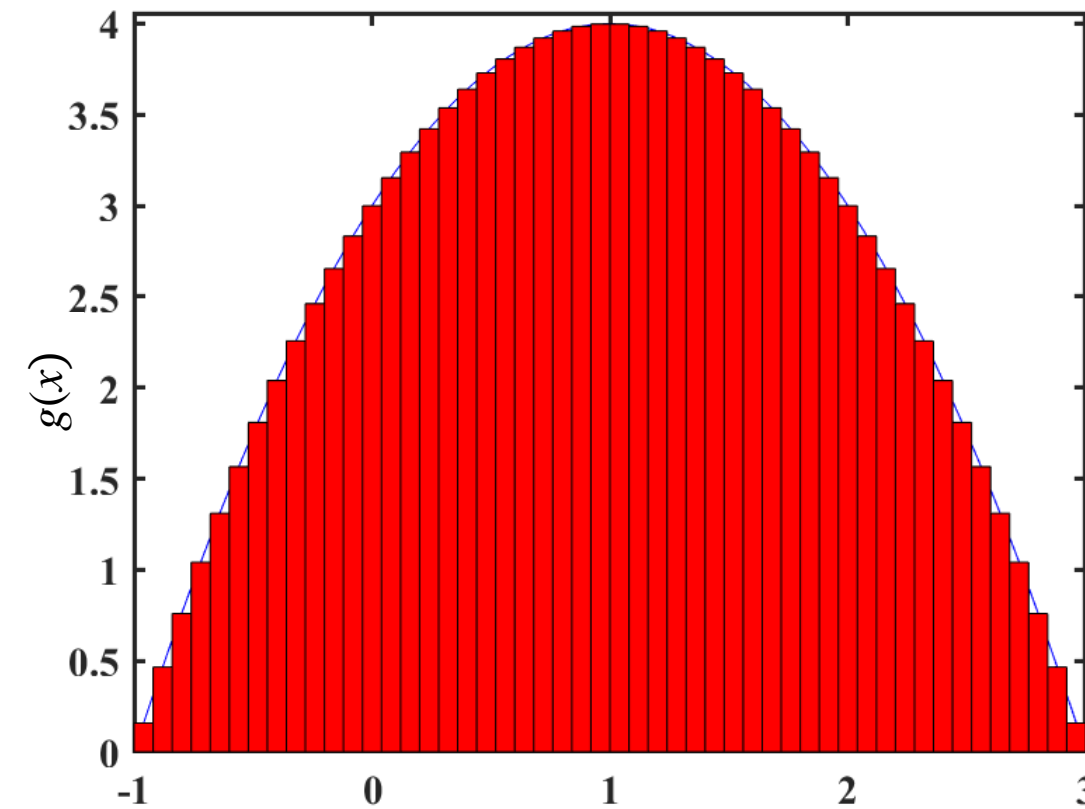

Deterministic Integration

goal

$$E(x) = \int_x g(x) dx$$

think of as

$$g(x) = xf(x)$$



$$g(x) = 4 - (x-1)^2 \quad x$$

← numerical

$$n=50, \quad \Delta x=0.080$$

$$\hat{E}(x) = \Delta x \sum_{i=1}^{50} g(x_i^*) = 10.6688$$

analytic

$$\int_{x=-1}^3 [4 - (x-1)^2] dx = 10.6667$$

% Numerical Integral

a=-1; b=3;

n=50; dx=(b-a)/n;

tpts=(a+dx/2:dx:b)';

gpts=4-(xpts-1).^2;

Exhat=dt*sum(gpts)

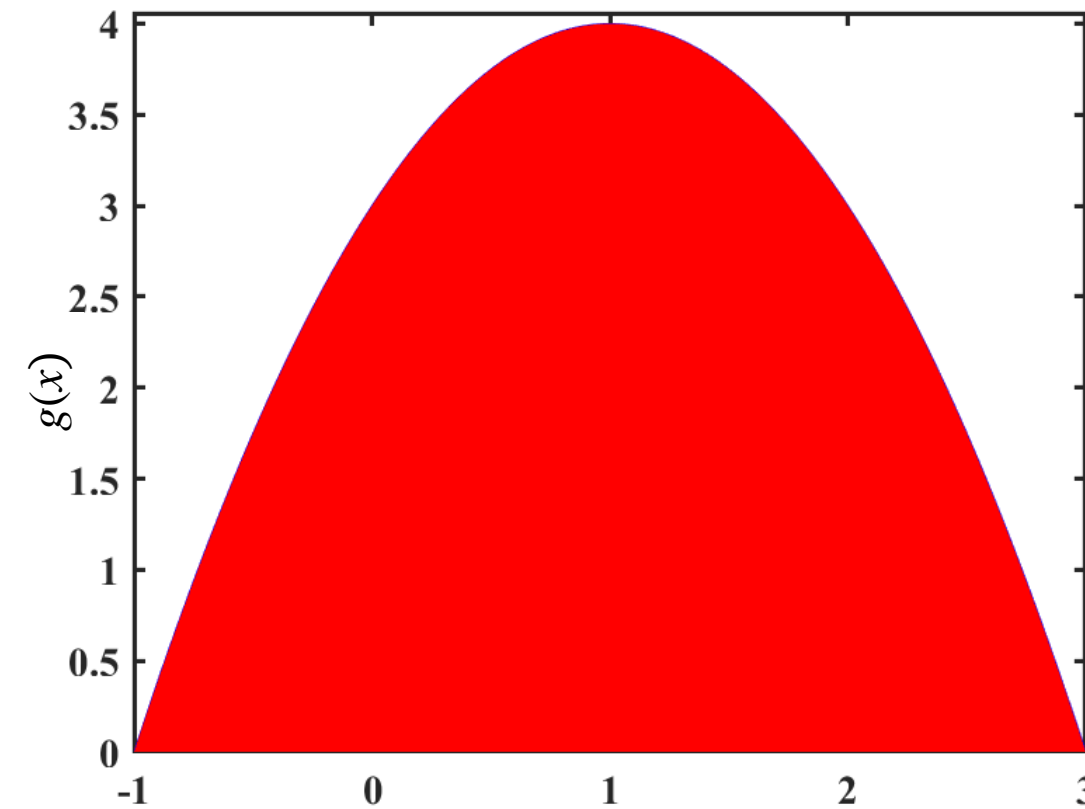
Deterministic Integration

goal

$$E(x) = \int_x g(x) dx$$

think of as

$$g(x) = xf(x)$$



$$g(x) = 4 - (x-1)^2 \quad x$$

← numerical

$$n=1000, \Delta x=0.004$$

$$\hat{E}(x) = \Delta x \sum_{i=1}^{1000} g(x_i^*) = 10.6667$$

analytic

$$\int_{x=-1}^3 [4 - (x-1)^2] dx = 10.6667$$

% Numerical Integral

a=-1; b=3;

n=1000; dx=(b-a)/n;

tpts=(a+dx/2:dx:b)';

gpts=4-(xpts-1).^2;

Exhat=dt*sum(gpts)

Chapter 3: Random Numbers

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Agenda

3.1 Pseudorandom Number Generation

3.2 Using Random numbers to Evaluate Integrals

3.1 Pseudorandom Number Generation - “Random” (0,1) numbers

Multiplicative Congruential Method

(1) Start with seed x_0

(2) Compute $x_n = ax_{n-1}$ modulo m , $n=0,1,2,\dots$
where a and m are given positive integers

choose $m = 2^{31}-1$ and $a=7^5$ for 32 bit machines

3.1 Pseudorandom Number Generation - “Random” (0,1) numbers

Mixed Congruential Method

(1) Start with seed x_0

(2) Compute $x_n = (ax_{n-1} + c) \text{ modulo } m$, $n=0,1,2,\dots$
where a , c , and m are given positive integers

choose $m =$ to be the computer's word length

3.2 Using Random numbers to Evaluate Integrals – Monte Carlo Integration

Let $g(x)$ be a function and suppose we want

$$\theta = \int_0^1 g(x) dx$$

If U is uniformly distributed over $(0,1)$, then

$$\theta = E(g(U))$$

Think of $f(u) = \begin{cases} 1 & \text{if } 0 \leq u \leq 1 \\ 0 & \text{otherwise} \end{cases}$

and

$$\theta = \int_0^1 g(u) f(u) du$$

3.2 Using Random numbers to Evaluate Integrals – Monte Carlo Integration

If we generate U_1, \dots, U_k independent uniforms

then $g(U_1), \dots, g(U_k)$ will be IID with mean $\theta = \int_0^1 g(x) dx$.

$$\sum_{i=1}^k \frac{g(U_i)}{k} \rightarrow E[g(U)] = \theta \quad \text{as } k \rightarrow \infty$$

$$\theta = \int_0^1 g(u) f(u) du$$

3.2 Using Random numbers to Evaluate Integrals – Monte Carlo Integration

If we want $\theta = \int_a^b g(x)dx$, then we can transform x to y as $y = (x - a)/(b - a)$, with $dy = dx/(b - a)$

$$\theta = \int_a^b g(x)dx \quad dx = (b - a)dy \quad \begin{array}{l} \text{limits } x = a \rightarrow y = 0 \\ x = b \rightarrow y = 1 \end{array}$$

$$x = a + y[b - a]$$

$$\theta = \int_0^1 g(a + [b - a]y)(b - a)dy$$

$$\theta = \int_0^1 h(y)dy \quad h(y) = g(a + [b - a]y)(b - a)$$

3.2 Using Random numbers to Evaluate Integrals – Monte Carlo Integration

Let's use this idea to evaluate the same integral.

$$(b - a) = 4$$

Deterministic

$$E(x|\cdot) = \int_x g(x)dx$$

$$\hat{E}_n(x|\cdot) = \Delta x \sum_{i=1}^n g(x_i^*)$$

$$\hat{E}_{10}(x|\cdot) = 10.7200$$

Stochastic

$$h(u) = g(a + (b - a)u)(b - a)$$

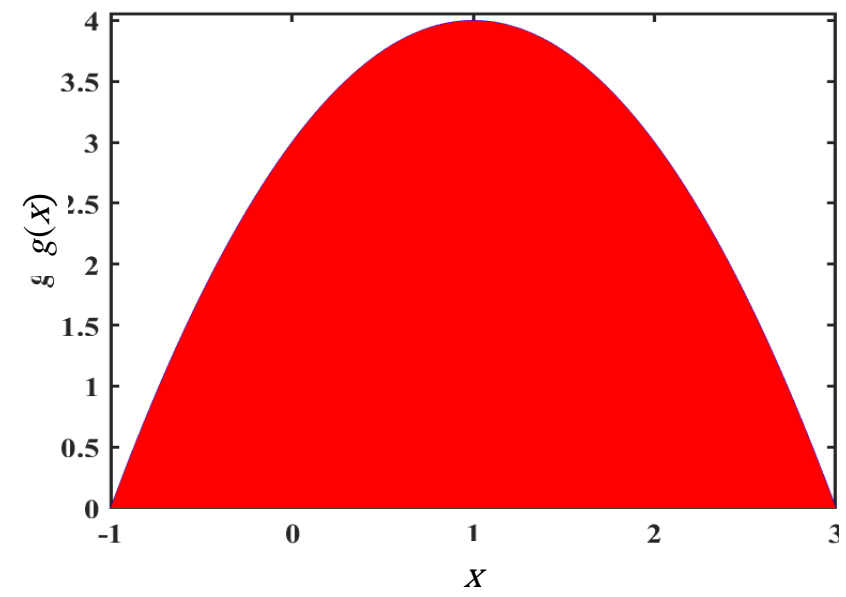
$$\frac{1}{n} \sum_{i=1}^n h(u_i) = 10.6718$$

```
% stochastic integral
rng('default')
a=-1; b=3; n=10^6;
u=rand(n,1);
t=(a+(b-a)*u);
hu=(4-(t-1).^2)*(b-a);
Exhat=sum(hu)/n
```

Theoretical

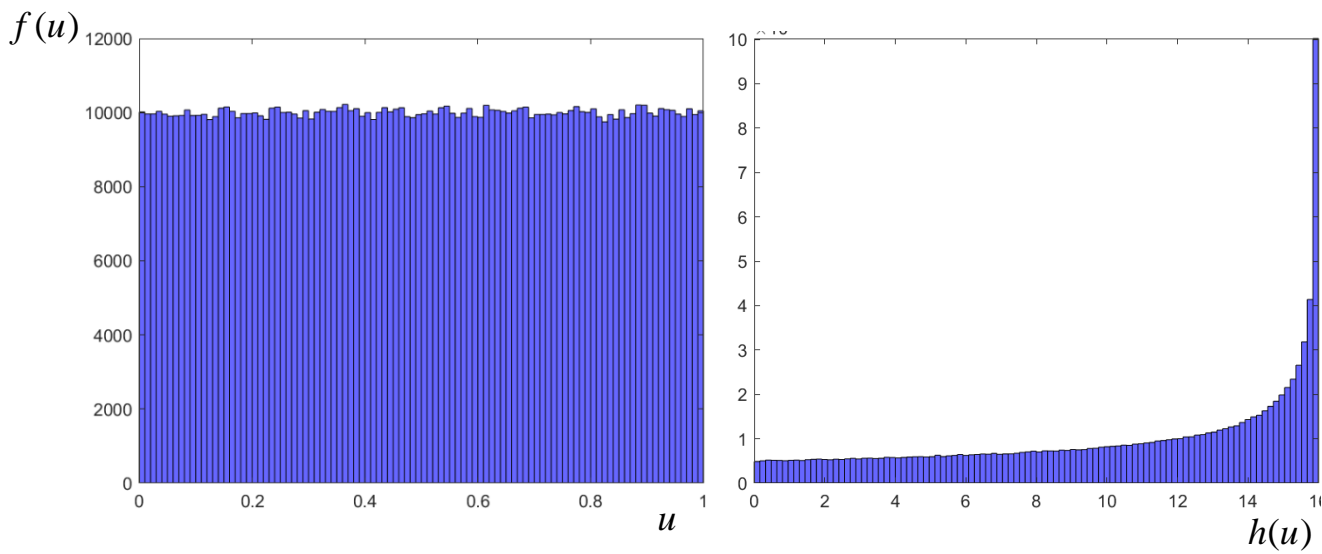
$$\int_{x=-1}^3 [4 - (x - 1)^2] dx = 10.6667$$

Deterministic



$$g(x) = 4 - (x - 1)^2$$

Stochastic



3.2 Using Random numbers to Evaluate Integrals – Monte Carlo Integration

Integrate $\theta = \int_0^1 2x \, dx$ using $U(0,1)$ random numbers.

Analytically, we'd get $\theta = \int_0^1 2x \, dx = 2 \frac{x^2}{2} \Big|_0^1 = 1$.

Or we can generate uniform numbers u_1, \dots, u_n and

Calculate $\theta \approx \frac{1}{n} \sum_{i=1}^n (2u_i)$.

Matlab Code:

```
rng('default')
```

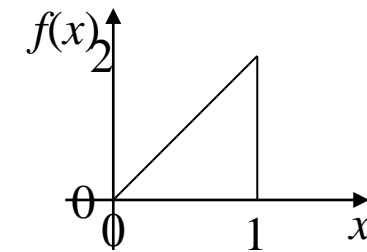
```
n=10^6;
```

```
u=rand(n,1);
```

```
theta=sum(2*u)/n
```

Matlab Output:

```
theta = 1.0006
```



3.2 Using Random numbers to Evaluate Integrals – Monte Carlo Integration

If we want $\theta = \int_0^{\infty} g(x)dx$, then we can transform x to y as $y = 1/(x + 1)$, with $dy = -dx/(x+1)^2 = -y^2 dx$

$$\theta = \int_0^{\infty} g(x)dx \quad \begin{array}{l} dx = -dy / y^2 \\ x = 1 / y - 1 \end{array} \quad \begin{array}{l} \text{limits } x = 0 \rightarrow y = 1 \\ x = \infty \rightarrow y = 0 \end{array}$$

$$\theta = \int_0^1 g(1/y - 1) dy / y^2$$

$$\theta = \int_0^1 h(y) dy \quad h(y) = g(1/y - 1) / y^2$$

3.2 Using Random numbers to Evaluate Integrals – Monte Carlo Integration

If we want $\theta = \int_{-\infty}^{+\infty} g(x)dx$, then we can transform x to y as $y = e^x / (1 + e^x)$, with $dy = dx e^x / (1 + e^x)^2 = y dx / (1 - y)$

$$\theta = \int_{-\infty}^{+\infty} g(x) dx \quad \begin{array}{l} dx = y dy / (1 - y) \\ x = \ln\left(\frac{y}{1-y}\right) \end{array} \quad \begin{array}{l} \text{limits } x = -\infty \rightarrow y = 0 \\ x = +\infty \rightarrow y = 1 \end{array}$$

$$\theta = \int_0^1 g\left(\ln\left(\frac{y}{1-y}\right)\right) \frac{dy}{y(1-y)} \quad y = e^x / (1 + e^x)$$

$$\theta = \int_0^1 h(y) dy \quad h(y) = g\left(\ln\left(\frac{y}{1-y}\right)\right) \frac{1}{y(1-y)}$$

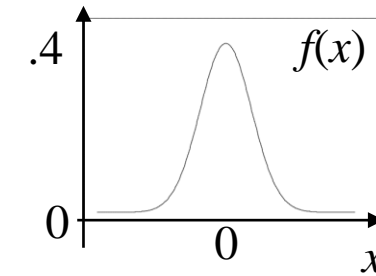
3.2 Using Random numbers to Evaluate Integrals – Monte Carlo Integration

Integrate $\theta = \int_{-\infty}^{+\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x^2} dx$ using $U(0,1)$ random numbers.

We know that $\theta = \int_{-\infty}^{+\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x^2} dx = 1$

Or we can generate uniform numbers u_1, \dots, u_n and

Calculate $\theta \approx \frac{1}{n} \sum_{i=1}^n \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2} \left(\ln \left(\frac{u_i}{1-u_i} \right) \right)^2} \frac{1}{u_i(1-u_i)} \approx 1$



$$\theta = \int_0^1 g \left(\ln \left(\frac{y}{1-y} \right) \right) \frac{dy}{y(1-y)}$$

3.2 Using Random numbers to Evaluate Integrals – Monte Carlo Integration

Most useful in evaluating multiple integrals.

$$\theta = \int_0^1 \int_0^1 \dots \int_0^1 g(x_1, \dots, x_n) dx_1 \dots dx_n$$

The key is to use $\theta = E[g(U_1, \dots, U_n)]$, where U_1, \dots, U_n are independent $U(0,1)$'s.

If we generate $U_1^{(1)}, \dots, U_n^{(1)}$ then $\theta = E[g(U_1, \dots, U_n)] \approx \sum_{i=1}^k \frac{g(U_1^i, \dots, U_n^i)}{k}$

$U_1^{(2)}, \dots, U_n^{(2)}$
 \vdots
 $U_1^{(k)}, \dots, U_n^{(k)}$

Homework 1

Problem A: $f(x) = \exp(\exp(x))$, $x \in [0,1]$

Integrate numerically with Matlab from $a=0$ to $b=1$.

Do by pencil and paper with $n=4$ intervals.

$$\Delta x = 0.25 \quad (x_1^*, x_2^*, x_3^*, x_4^*) = (0.125, 0.375, 0.625, 0.875)$$

Write a Matlab program to repeat with $n=4$ intervals.

Change to $n=100$. Compare results.

Homework 1

Chapter 3: # 1*, 3#, 7, 9, 11.

*Repeat generating 10^4 of these.

Compute mean, variance, and make a histogram.

Repeat using Matlab's `rand()` command.

Compare results from to Matlab's `rand()`.

#Compare results to numerical integration in Problem A with $n=100$.

Homework 1

Problem B:

Evaluate the double integral

$$\iint x_1 f(x_1, x_2) dx_1 dx_2 \quad \text{where} \quad x_1 \in \mathbb{R}, x_2 \in \mathbb{R}$$

for PDF $f(x_1, x_2) = \frac{1}{\sqrt{2\pi\sigma_1^2}} e^{-\frac{1}{2\sigma_1^2}(x_1-\mu_1)^2} \frac{1}{\sqrt{2\pi\sigma_2^2}} e^{-\frac{1}{2\sigma_2^2}(x_2-\mu_2)^2}$.

$$\mu_1 = 10$$

$$\mu_2 = 5$$

$$\sigma_1^2 = 4$$

$$\sigma_2^2 = 9$$

a. Use numerical integration by breaking up (x_1, x_2)

into rectangles. Use $\Delta x_1 \Delta x_2 \sum_{x_2} \sum_{x_1} x_1 f(x_1, x_2)$.

b. Use stochastic integration $\sum_{i=1}^k \frac{g(U_1^i, U_2^i)}{k}$ after transforming (x_1, x_2) to (u_1, u_2) .

c. Pencil and paper integration?

