

*Numerical Methods in Economics*  
MIT Press, 1998

## Notes for Chapter 7: Numerical Integration and Differentiation

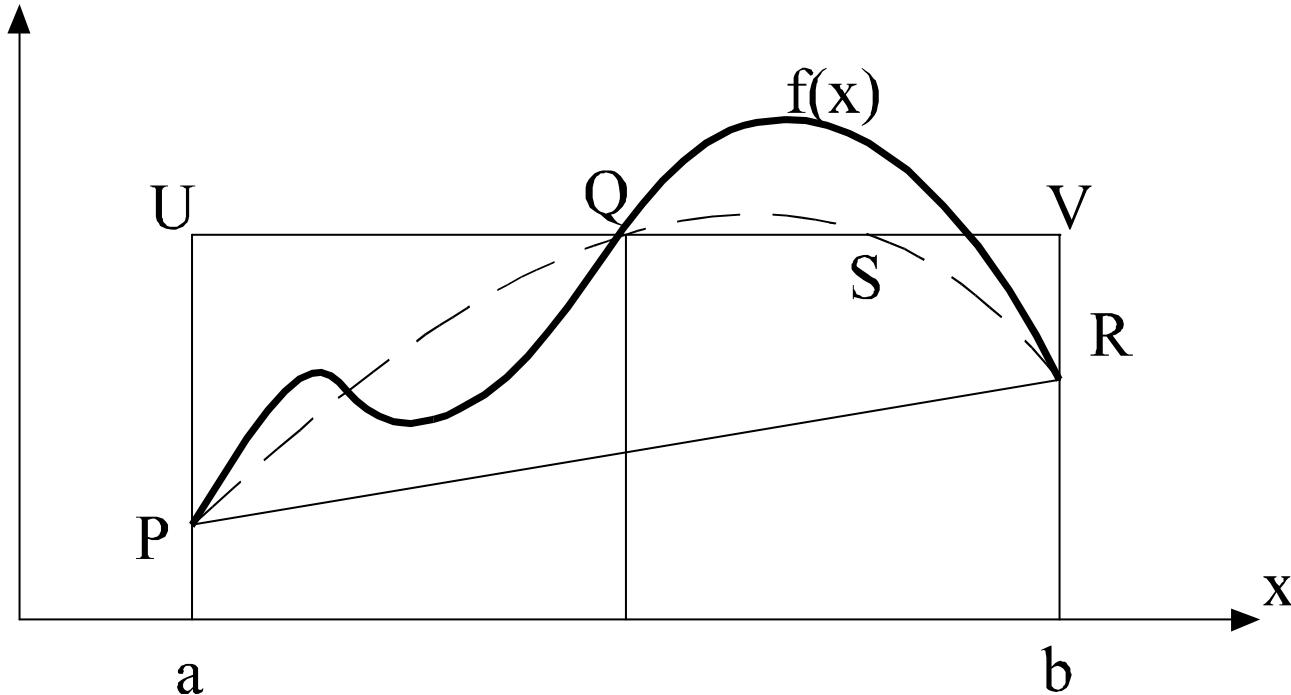
October 29, 2008

# Integration

- Most integrals cannot be evaluated analytically
- Integrals frequently arise in economics
  - Expected utility
  - Discounted utility and profits over a long horizon
  - Bayesian posterior
  - Likelihood functions
  - Solution methods for dynamic economic models

## Newton-Cotes Formulas

- Idea: Approximate function with low order polynomials and then integrate approximation



- Step function approximation:

- Compute constant function equalling  $f(x)$  at midpoint of  $[a, b]$
- Integral approximation is  $aUQVb$  box

- Linear function approximation:

- Compute linear function interpolating  $f(x)$  at  $a$  and  $b$
- Integral approximation is trapezoid  $aPRb$

- Parabolic function approximation:
  - Compute parabola interpolating  $f(x)$  at  $a$ ,  $b$ , and  $(a + b)/2$
  - Integral approximation is area of  $aPQRb$

- Midpoint Rule: piecewise step function approximation

$$\int_a^b f(x) \, dx = (b - a) f\left(\frac{a + b}{2}\right) + \frac{(b - a)^3}{24} f''(\xi)$$

- Simple rule: for some  $\xi \in [a, b]$

$$\int_a^b f(x) \, dx = (b - a) f\left(\frac{a + b}{2}\right) + \frac{(b - a)^3}{24} f''(\xi)$$

- Composite midpoint rule:

\* nodes:  $x_j = a + (j - \frac{1}{2})h$ ,  $j = 1, 2, \dots, n$ ,  $h = (b - a)/n$

\* for some  $\xi \in [a, b]$

$$\int_a^b f(x) \, dx = h \sum_{j=1}^n f\left(a + (j - \frac{1}{2})h\right) + \frac{h^2(b - a)}{24} f''(\xi)$$

- Trapezoid Rule: piecewise linear approximation

- Simple rule: for some  $\xi \in [a, b]$

$$\int_a^b f(x) dx = \frac{b-a}{2} [f(a) + f(b)] - \frac{(b-a)^3}{12} f''(\xi)$$

- Composite trapezoid rule:

\* nodes:  $x_j = a + (j - \frac{1}{2})h, j = 1, 2, \dots, n, h = (b - a)/n$

\* for some  $\xi \in [a, b]$

$$\begin{aligned} \int_a^b f(x) dx &= \frac{h}{2} [f_0 + 2f_1 + \cdots + 2f_{n-1} + f_n] \\ &\quad - \frac{h^2 (b-a)}{12} f''(\xi) \end{aligned}$$

- Simpson's Rule: piecewise quadratic approximation

- for some  $\xi \in [a, b]$

$$\int_a^b f(x) \, dx = \left( \frac{b-a}{6} \right) \left[ f(a) + 4f\left(\frac{a+b}{2}\right) + f(b) \right] - \frac{(b-a)^5}{2880} f^{(4)}(\xi)$$

- Composite Simpson's rule: for some  $\xi \in [a, b]$

$$\int_a^b f(x) \, dx = \frac{h}{3} [f_0 + 4f_1 + 2f_2 + 4f_3 + \cdots + 4f_{n-1} + f_n] - \frac{h^4(b-a)}{180} f^{(4)}(\xi)$$

- Obscure rules for degree 3, 4, etc. approximations.

## Change of Variables Formula and Infinite Domains

- Problem: How do we approximate integrals with infinite domains?

$$\int_0^\infty f(x)dx \equiv \lim_{b \rightarrow \infty} \int_0^b f(x) dx$$

- Truncation (a bad idea): For large  $b$ , use

$$\int_0^\infty f(x)dx \doteq \int_0^b f(x) dx$$

- Change of variables theorem:

**Theorem 1** *If  $\phi : \mathbb{R} \rightarrow \mathbb{R}$  is a monotonically increasing,  $C^1$  function on the (possibly infinite) interval  $[a, b]$ , then for any integrable  $g(y)$  on  $[a, b]$ ,*

$$\int_a^b g(y) dy = \int_{\phi^{-1}(a)}^{\phi^{-1}(b)} g(\phi(x)) \phi'(x) dx. \quad (7.1.8)$$

- COV Objective: find a  $x(z)$  function such that

$$\int_0^\infty f(x) dx = \int_0^1 f(x(z)) x'(z) dz$$

can be accurately computed.

–  $x : (0, 1) \rightarrow (0, \infty)$ :  $x(z) = \frac{z}{1-z}$ ,  $x'(z) = \frac{1}{(1-z)^2}$

\* Example:

$$\int_0^\infty e^{-t} t^n dt = \int_0^1 e^{-z/(1-z)} \left(\frac{z}{1-z}\right)^n (1-z)^{-2} dz.$$

\* All derivatives are bounded, so Newton-Cotes error bound formulas applies.

–  $x : (0, 1) \rightarrow (-\infty, \infty)$ :  $x(z) = \ln\left(\frac{z}{1-z}\right)$ ,  $x'(z) = (z(1-z))^{-1}$

\* Example:  $\int_{-\infty}^\infty e^{-x^2} f(x) dx$  becomes

$$\begin{aligned} & \int_0^1 e^{-(\ln(\frac{z}{1-z}))^2} f\left(\ln(\frac{z}{1-z})\right) \frac{dz}{(1-z)z} \\ &= \int_0^1 \left(\frac{1-z}{z}\right)^{\ln(\frac{z}{1-z})} f\left(\ln(\frac{z}{1-z})\right) \frac{dz}{z(1-z)} \end{aligned}$$

\* Integrand's derivatives are bounded if  $f$  is exponentially bounded

– Bad COV formula

\*  $x(z) = \left(\ln\frac{z}{1-z}\right)^{1/3}$  maps  $(0, 1)$  onto  $(-\infty, \infty)$

\* Application to  $\int_{-\infty}^\infty f(x) e^{-x^2} dx$  often creates an integrand with unbounded derivatives.

## Gaussian Formulas

- All integration formulas are of form

$$\int_a^b f(x) dx \doteq \sum_{i=1}^n \omega_i f(x_i) \quad (7.2.1)$$

for some *quadrature nodes*  $x_i \in [a, b]$  and *quadrature weights*  $\omega_i$ .

- Newton-Cotes use arbitrary  $x_i$
  - Gaussian quadrature uses good choices of  $x_i$  nodes and  $\omega_i$  weights.
- Exact quadrature formulas:
    - Let  $\mathcal{F}_k$  be the space of degree  $k$  polynomials
    - A quadrature formula is exact of degree  $k$  if it correctly integrates each function in  $\mathcal{F}_k$
    - Gaussian quadrature formulas use  $n$  points and are exact of degree  $2n - 1$

**Theorem 2** Suppose that  $\{\varphi_k(x)\}_{k=0}^{\infty}$  is an orthonormal family of polynomials with respect to  $w(x)$  on  $[a, b]$ .

1. Define  $q_k$  so that  $\varphi_k(x) = q_k x^k + \cdots$ .
2. Let  $x_i$ ,  $i = 1, \dots, n$  be the  $n$  zeros of  $\varphi_n(x)$
3. Let  $\omega_i = -\frac{q_{n+1}/q_n}{\varphi'_n(x_i) \varphi_{n+1}(x_i)} > 0$

Then

1.  $a < x_1 < x_2 < \cdots < x_n < b$ ;
2. if  $f \in C^{(2n)}[a, b]$ , then for some  $\xi \in [a, b]$ ,

$$\int_a^b w(x) f(x) dx = \sum_{i=1}^n \omega_i f(x_i) + \frac{f^{(2n)}(\xi)}{q_n^2 (2n)!};$$

3. and  $\sum_{i=1}^n \omega_i f(x_i)$  is the unique formula on  $n$  nodes that exactly integrates  $\int_a^b f(x) w(x) dx$  for all polynomials in  $\mathcal{F}_{2n-1}$ .

## Gauss-Chebyshev Quadrature

- Domain:  $[-1, 1]$
- Weight:  $(1 - x^2)^{-1/2}$
- Formula:

$$\int_{-1}^1 f(x)(1 - x^2)^{-1/2} dx = \frac{\pi}{n} \sum_{i=1}^n f(x_i) + \frac{\pi}{2^{2n-1}} \frac{f^{(2n)}(\xi)}{(2n)!} \quad (7.2.4)$$

for some  $\xi \in [-1, 1]$ , with quadrature nodes

$$x_i = \cos\left(\frac{2i-1}{2n}\pi\right), \quad i = 1, \dots, n. \quad (7.2.5)$$

## Arbitrary Domains

- Want to approximate  $\int_a^b f(x) dx$ 
  - Different range, no weight function
  - Linear change of variables  $x = -1 + 2(y - a)(b - a)$
  - Multiply the integrand by  $(1 - x^2)^{1/2} / (1 - x^2)^{1/2}$ .
  - C.O.V. formula

$$\int_a^b f(y) dy = \frac{b-a}{2} \int_{-1}^1 f\left(\frac{(x+1)(b-a)}{2} + a\right) \frac{(1-x^2)^{1/2}}{(1-x^2)^{1/2}} dx$$

- Gauss-Chebyshev quadrature produces

$$\int_a^b f(y) dy \doteq \frac{\pi(b-a)}{2n} \sum_{i=1}^n f\left(\frac{(x_i+1)(b-a)}{2} + a\right) (1-x_i^2)^{1/2}$$

where the  $x_i$  are Gauss-Chebyshev nodes over  $[-1, 1]$ .

## Gauss-Legendre Quadrature

- Domain:  $[-1, 1]$

- Weight: 1

- Formula:

$$\int_{-1}^1 f(x) dx = \sum_{i=1}^n \omega_i f(x_i) + \frac{2^{2n+1}(n!)^4}{(2n+1)!(2n)!} \cdot \frac{f^{(2n)}(\xi)}{(2n)!}$$

for some  $-1 \leq \xi \leq 1$ .

- Convergence:

– use  $n! \doteq e^{-n-1} n^{n+1/2} \sqrt{2\pi n}$

– error bounded above by  $\pi 4^{-n} M$

$$M = \sup_m \left[ \max_{-1 \leq x \leq 1} \frac{f^{(m)}(x)}{m!} \right]$$

– Exponential convergence for analytic functions

- In general,

$$\int_a^b f(x) dx \doteq \frac{b-a}{2} \sum_{i=1}^n \omega_i f \left( \frac{(x_i+1)(b-a)}{2} + a \right)$$

- Use values for Gaussian nodes and weights from tables instead of programs; tables will have 16 digit accuracy

Table 7.2: Gauss – Legendre Quadrature

$N$	$x_i$	$\omega_i$
2	0.5773502691	0.1000000000(1)
3	0.7745966692	0.5555555555
	0.0000000000	0.8888888888
5	0.9061798459	0.2369268850
	0.5384693101	0.4786286704
	0.0000000000	0.5688888888
10	0.9739065285	0.6667134430(-1)
	0.8650633666	0.1494513491
	0.6794095682	0.2190863625
	0.4333953941	0.2692667193
	0.1488743389	0.2955242247

## Life-cycle example:

- $c(t) = 1 + t/5 - 7(t/50)^2$ , where  $0 \leq t \leq 50$ .
- Discounted utility is  $\int_0^{50} e^{-\rho t} u(c(t)) dt$
- $\rho = 0.05$ ,  $u(c) = c^{1+\gamma}/(1+\gamma)$ .
- Errors in computing  $\int_0^{50} e^{-.05t} \left(1 + \frac{t}{5} - 7\left(\frac{t}{50}\right)^2\right)^{1-\gamma} dt$

$\gamma =$	.5	1.1	3	10
Truth	1.24431	.664537	.149431	.0246177
Rule:	GLeg 3	5(-3)	2(-3)	3(-2)
	GLeg 5	1(-4)	8(-5)	5(-3)
	GLeg 10	1(-7)	1(-7)	2(-5)
	GLeg 15	1(-10)	2(-10)	9(-8)
	GLeg 20	7(-13)	9(-13)	3(-10)
				6(-7)

## Gauss-Hermite Quadrature

- Domain:  $[-\infty, \infty]$

- Weight:  $e^{-x^2}$

- Formula:

$$\int_{-\infty}^{\infty} f(x)e^{-x^2} dx = \sum_{i=1}^n \omega_i f(x_i) + \frac{n! \sqrt{\pi}}{2^n} \cdot \frac{f^{(2n)}(\xi)}{(2n)!}$$

for some  $\xi \in (-\infty, \infty)$ .

Table 7.4: Gauss – Hermite Quadrature

$N$	$x_i$	$\omega_i$
2	0.7071067811	0.8862269254
3	0.1224744871(1)	0.2954089751
	0.0000000000	0.1181635900(1)
4	0.1650680123(1)	0.8131283544(−1)
	0.5246476232	0.8049140900
7	0.2651961356(1)	0.9717812450(−3)
	0.1673551628(1)	0.5451558281(−1)
	0.8162878828	0.4256072526
	0.0000000000	0.8102646175
10	0.3436159118(1)	0.7640432855(−5)
	0.2532731674(1)	0.1343645746(−2)
	0.1756683649(1)	0.3387439445(−1)
	0.1036610829(1)	0.2401386110
	0.3429013272	0.6108626337

- Normal Random Variables

- $Y$  is distributed  $N(\mu, \sigma^2)$
- Expectation is integration:

$$E\{f(Y)\} = (2\pi\sigma^2)^{-1/2} \int_{-\infty}^{\infty} f(y) e^{-\frac{(y-\mu)^2}{2\sigma^2}} dy$$

- Use Gauss-Hermite quadrature
  - \* linear COV  $x = (y - \mu)/\sqrt{2}\sigma$
  - \* COV formula:

$$\int_{-\infty}^{\infty} f(y) e^{-(y-\mu)^2/(2\sigma^2)} dy = \int_{-\infty}^{\infty} f(\sqrt{2}\sigma x + \mu) e^{-x^2} \sqrt{2}\sigma dx$$

- \* COV quadrature formula:

$$E\{f(Y)\} \doteq \pi^{-\frac{1}{2}} \sum_{i=1}^n \omega_i f(\sqrt{2}\sigma x_i + \mu)$$

where the  $\omega_i$  and  $x_i$  are the Gauss-Hermite quadrature weights and nodes over  $[-\infty, \infty]$ .

- Portfolio example
  - An investor holds one bond which will be worth 1 in the future and equity whose value is  $Z$ , where  $\ln Z \sim \mathcal{N}(\mu, \sigma^2)$ .
  - Expected utility is

$$U = (2\pi\sigma^2)^{-1/2} \int_{-\infty}^{\infty} u(1 + e^z) e^{-(z-\mu)^2/2\sigma^2} dz \quad (7.2.12)$$

$$u(c) = \frac{c^{1+\gamma}}{1+\gamma}$$

and the certainty equivalent of (7.2.12) is  $u^{-1}(U)$ .

- Errors in certainty equivalents: Table 7.5

Rule	$\gamma:$	-.5	-1.1	-2.0	-5.0	-10.0
GH2		1(-4)	2(-4)	3(-4)	6(-3)	3(-2)
GH3		1(-6)	3(-6)	9(-7)	7(-5)	9(-5)
GH4		2(-8)	7(-8)	4(-7)	7(-6)	1(-4)
GH7		3(-10)	2(-10)	3(-11)	3(-9)	1(-9)
GH13		3(-10)	2(-10)	3(-11)	5(-14)	2(-13)

- The certainty equivalent of (7.2.12) with  $\mu = 0.15$  and  $\sigma = 0.25$  is 2.34. So, relative errors are roughly the same.

## Gauss-Laguerre Quadrature

- Domain:  $[0, \infty]$

- Weight:  $e^{-x}$

- Formula:

$$\int_0^\infty f(x)e^{-x}dx = \sum_{i=1}^n \omega_i f(x_i) + (n!)^2 \frac{f^{(2n)}(\xi)}{(2n)!}$$

for some  $\xi \in [0, \infty)$ .

- General integral

- Linear COV  $x = r(y - a)$
- COV formula

$$\int_a^\infty e^{-ry} f(y) dy \doteq \frac{e^{-ra}}{r} \sum_{i=1}^n \omega_i f\left(\frac{x_i}{r} + a\right)$$

where the  $\omega_i$  and  $x_i$  are the Gauss-Laguerre quadrature weights and nodes over  $[0, \infty]$ .

Table 7.6: Gauss – Laguerre Quadrature

$N$	$x_i$	$\omega_i$
2	0.5857864376	0.8535533905
	0.3414213562(1)	0.1464466094
3	0.4157745567	0.7110930099
	0.2294280360(1)	0.2785177335
	0.6289945082(1)	0.1038925650(–1)
4	0.3225476896	0.6031541043
	0.1745761101(1)	0.3574186924
	0.4536620296(1)	0.3888790851(–1)
	0.9395070912(1)	0.5392947055(–3)
7	0.1930436765	0.4093189517
	0.1026664895(1)	0.4218312778
	0.2567876744(1)	0.1471263486
	0.4900353084(1)	0.2063351446(–1)
	0.8182153444(1)	0.1074010143(–2)
	0.1273418029(2)	0.1586546434(–4)
	0.1939572786(2)	0.3170315478(–7)

- Present Value Example

- Use Gauss-Laguerre quadrature to compute present values.
- Suppose discounted profits equal

$$\eta \left( \frac{\eta - 1}{\eta} \right)^{\eta-1} \int_0^{\infty} e^{-rt} m(t)^{1-\eta} dt.$$

- Errors: Table 7.7

	$r = .05$	$r = .10$	$r = .05$
	$\lambda = .05$	$\lambda = .05$	$\lambda = .20$
Truth:	49.7472	20.3923	74.4005
Errors: GLag 4	3(-1)	4(-2)	6(0)
GLag 5	7(-3)	7(-4)	3(0)
GLag 10	3(-3)	6(-5)	2(-1)
GLag 15	6(-5)	3(-7)	6(-2)
GLag 20	3(-6)	8(-9)	1(-2)

- Gauss-Laguerre integration implicitly assumes that  $m(t)^{1-\eta}$  is a polynomial.
  - \* When  $\lambda = 0.05$ ,  $m(t)$  is nearly constant
  - \* When  $\lambda = 0.20$ ,  $m(t)^{1-\eta}$  is less polynomial-like.

## Do-It-Yourself Gaussian Formulas

- Question: What should you do if your problem does not fit one of the conventional integral problems?
- Answer: Create your own Gaussian formula!
- Theorem: Let  $w(x)$  be a weight function on  $[a, b]$ , and suppose that all moments exist; i.e.,  $\int_a^b x^i w(x) dx < \infty$  for all  $i$ . Then for all  $n$  there exists *quadrature nodes*  $x_i \in [a, b]$  and *quadrature weights*  $\omega_i$  such that

$$\int_a^b w(x) f(x) dx = \sum_{i=1}^n \omega_i f(x_i)$$

is exact for all degree  $2n - 1$  polynomials. The nodes are the zeros of the monic polynomial  $p(x) = x^n + a_{n-1}x^{n-1} + a_{n-2}x^{n-2} \dots + a_0$  that minimizes

$$\int_a^b p(x)^2 w(x) dx$$

and the weights are chosen to satisfy the overdetermined linear equations

$$\int_a^b x^k w(x) dx = \sum_{i=1}^n \omega_i x_i^k, \quad k = 0, 1, \dots, 2n - 1$$

- Implementation: The optimization problem for finding the  $a_i$  coefficients is a smooth unconstrained optimization problem and the equations for finding the weights  $\omega_i$  are linear, but both are ill-conditioned. However, using extended precision arithmetic (as is possible in Mathematica) will solve those problems.

## General Applicability of Gaussian Quadrature

**Theorem 3** (*Gaussian quadrature convergence*) *If  $f$  is Riemann Integrable on  $[a, b]$ , the error in the  $n$ -point Gauss-Legendre rule applied to  $\int_a^b f(x) dx$  goes to 0 as  $n \rightarrow \infty$ .*

**Comparisons with Newton-Cotes formulas:** Table 7.1

Rule	$n$	$\int_0^1 x^{1/4} dx$	$\int_1^{10} x^{-2} dx$	$\int_0^1 e^x dx$	$\int_1^{-1} (x + .05)^+ dx$
Trapezoid	4	0.7212	1.7637	1.7342	0.6056
	7	0.7664	1.1922	1.7223	0.5583
	10	0.7797	1.0448	1.7200	0.5562
	13	0.7858	0.9857	1.7193	0.5542
Simpson	3	0.6496	1.3008	1.4662	0.4037
	7	0.7816	1.0017	1.7183	0.5426
	11	0.7524	0.9338	1.6232	0.4844
	15	0.7922	0.9169	1.7183	0.5528
G-Legendre	4	0.8023	0.8563	1.7183	0.5713
	7	0.8006	0.8985	1.7183	0.5457
	10	0.8003	0.9000	1.7183	0.5538
	13	0.8001	0.9000	1.7183	0.5513
Truth		.80000	.90000	1.7183	0.55125

## Multidimensional Integration

- Most economic problems have several dimensions
  - Multiple assets
  - Multiple error terms
- Multidimensional integrals are much more difficult
  - Simple methods suffer from curse of dimensionality
  - There are methods which avoid curse of dimensionality

## Product Rules

- Build product rules from one-dimension rules
- Let  $x_i^\ell, \omega_i^\ell, i = 1, \dots, m$ , be one-dimensional quadrature points and weights in dimension  $\ell$  from a Newton-Cotes rule or the Gauss-Legendre rule.
- The *product rule*

$$\int_{[-1,1]^d} f(x) dx \doteq \sum_{i_1=1}^m \cdots \sum_{i_d=1}^m \omega_{i_1}^1 \omega_{i_2}^2 \cdots \omega_{i_d}^d f(x_{i_1}^1, x_{i_2}^2, \dots, x_{i_d}^d)$$

- Gaussian structure prevails
  - Suppose  $w^\ell(x)$  is weighting function in dimension  $\ell$
  - Define the  $d$ -dimensional weighting function.

$$W(x) \equiv W(x_1, \dots, x_d) = \prod_{\ell=1}^d w^\ell(x_\ell)$$

- Product Gaussian rules are based on product orthogonal polynomials.
- Curse of dimensionality:
  - $m^d$  functional evaluations is  $m^d$  for a  $d$ -dimensional problem with  $m$  points in each direction.
  - Problem worse for Newton-Cotes rules which are less accurate in  $\mathbb{R}^1$ .

## Monomial Formulas: A Nonproduct Approach

- Method
- Choose  $x^i \in D \subset \mathbb{R}^d$ ,  $i = 1, \dots, N$
- Choose  $\omega_i \in \mathbb{R}$ ,  $i = 1, \dots, N$
- Quadrature formula

$$\int_D f(x) dx \doteq \sum_{i=1}^N \omega_i f(x^i) \quad (7.5.3)$$

- A monomial formula is complete for degree  $\ell$  if

$$\sum_{i=1}^N \omega_i p(x^i) = \int_D p(x) dx \quad (7.5.3)$$

for all polynomials  $p(x)$  of total degree  $\ell$ ; recall that  $\mathcal{P}_\ell$  was defined in chapter 6 to be the set of such polynomials.

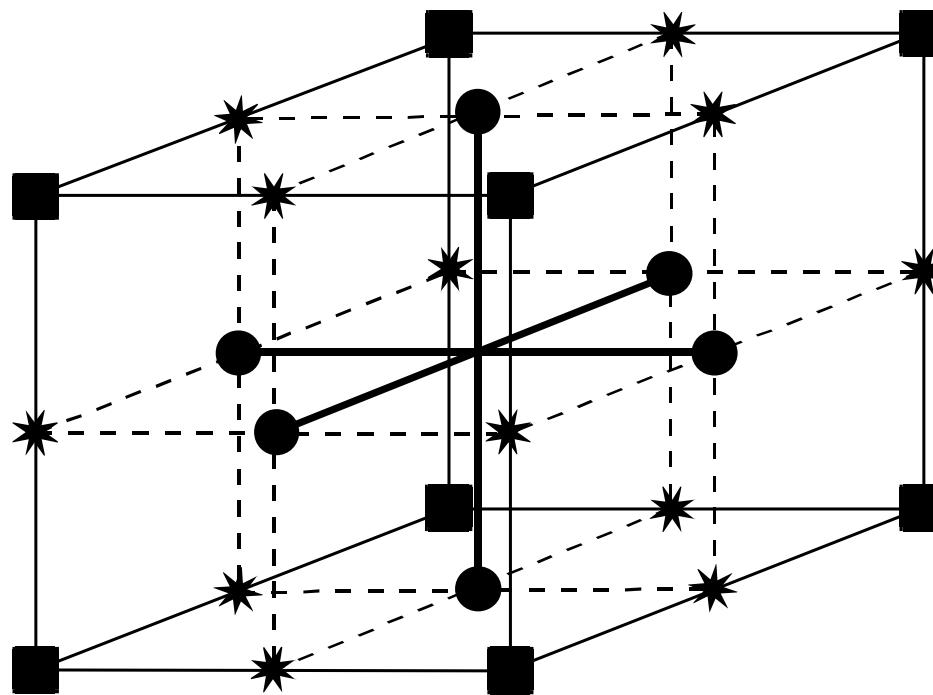
- For the case  $\ell = 2$ , this implies the equations

$$\begin{aligned} \sum_{i=1}^N \omega_i &= \int_D 1 \cdot dx \\ \sum_{i=1}^N \omega_i x_j^i &= \int_D x_j dx, \quad j = 1, \dots, d \\ \sum_{i=1}^N \omega_i x_j^i x_k^i &= \int_D x_j x_k dx, \quad j, k = 1, \dots, d \end{aligned} \quad (7.5.4)$$

–  $1 + d + \frac{1}{2}d(d + 1)$  equations

–  $N$  weights  $\omega_i$  and the  $N$  nodes  $x^i$  each with  $d$  components, yielding a total of  $(d + 1)N$  unknowns.

# Quadrature Node Sets



- Natural types of nodes:
  - The center
  - The circles: centers of faces
  - The stars: centers of edges
  - The squares: vertices

- Some monomial formulas will take some combinations of these sets
- Other types of collections are possible

- Simple examples

- Let  $e^j \equiv (0, \dots, 1, \dots, 0)$  where the ‘1’ appears in column  $j$ .
- $2d$  points and exactly integrates all elements of  $\mathcal{P}_3$  over  $[-1, 1]^d$

$$\int_{[-1,1]^d} f \doteq \omega \sum_{i=1}^d (f(ue^i) + f(-ue^i))$$

$$u = \left(\frac{d}{3}\right)^{1/2}, \quad \omega = \frac{2^{d-1}}{d}$$

- For  $\mathcal{P}_5$  the following scheme works:

$$\int_{[-1,1]^d} f \doteq \omega_1 f(0) + \omega_2 \sum_{i=1}^d (f(ue^i) + f(-ue^i))$$

$$+ \omega_3 \sum_{\substack{1 \leq i < d, \\ i < j \leq d}} (f(u(e^i \pm e^j)) + f(-u(e^i \pm e^j)))$$

where

$$\omega_1 = 2^d(25d^2 - 115d + 162), \quad \omega_2 = 2^d(70 - 25d)$$

$$\omega_3 = \frac{25}{324} 2^d, \quad u = \left(\frac{3}{5}\right)^{1/2}.$$

- Smolyak (a.k.a., sparse) grids (see pictures on next slide)

## Existence Result for Monomial Formulas

**Theorem 4 (Mysovskikh)** Let  $w(x)$  be a nonnegative weighting function on  $D \subset \mathbb{R}^d$  such that each moment

$$\int_D w(x) x_1^{i_1} \cdots x_d^{i_d} dx_1 \cdots dx_d$$

exists for  $i_1, \dots, i_d \geq 0$ ,  $i_1 + \cdots + i_d \leq m$ . Then, for some  $N \leq (m+d)!/(m!d!)$ , there exists  $N$  positive weights,  $\omega_i$ , and  $N$  nodes,  $x^i$ , such that for each multi-index  $|\alpha| \leq m$ ,

$$\int_D w(x) x^\alpha dx = \sum_{i=1}^N \omega_i (x^i)^\alpha.$$

- Purely existential
- Solving equations is difficult. However, recent advances in solving polynomial systems make this feasible for small problems, and possibly even for moderately large systems when one imposes symmetry conditions
- Formulas do not suffer from curse of dimensionality
- See Stroud and Secrest book for a large list of formulas.

## Numerical Differentiation

- One-sided formula

$$f'(x) \doteq \frac{f(x+h) - f(x)}{h} \quad (7.7.1)$$

- What should  $h$  be in light of computer errors?
- Error Analysis when  $\hat{f}$  is computer version of  $f$

- Suppose  $|f(x) - \hat{f}(x)| \leq \varepsilon$ ,
- Actual machine approximation is

$$D(h) = \frac{\hat{f}(x+h) - \hat{f}(x)}{h}$$

- Error bound is

$$\left| D(h) - \frac{f(x+h) - f(x)}{h} \right| \leq \frac{2\varepsilon}{h}$$

- Taylor's theorem: for some  $\xi \in [x, x+h]$

$$f'(x) = \frac{f(x+h) - f(x)}{h} - \frac{h}{2} f''(\xi)$$

- If  $M_2 > 0$  is an upper bound on  $|f''|$  near  $x$ , then error bound is

$$|f'(x) - D(h)| \leq \frac{2\varepsilon}{h} + \frac{h}{2} M_2 \quad (7.7.2)$$

- Upper bound on error is minimized at

$$h^* = 2\sqrt{\frac{\varepsilon}{M_2}} \quad (7.7.3)$$

- The upper bound on error equals  $2\sqrt{\varepsilon M_2}$ .

- Two-Sided Difference Formula

- Two-sided formula

$$f'(x) \doteq \frac{f(x+h) - f(x-h)}{2h} \quad (7.7.4)$$

- Error is  $\frac{h^2}{6} f'''(\xi)$  for some  $\xi \in [x-h, x+h]$ .
- Round-off error of the approximation error is  $\varepsilon/h$
- Total error of

$$\frac{M_3 h^2}{6} + \frac{\varepsilon}{h}$$

if  $M_3 > |f'''|$  near  $x$ .

- Optimal  $h$  is  $\frac{3\varepsilon}{M_3}^{1/3}$  with error upper bound of  $2\varepsilon^{2/3} M_3^{1/3} 9^{1/3}$ .
- Two-sided formula reduced error from order  $\varepsilon^{1/2}$  to order  $\varepsilon^{2/3}$ .
- On a twelve-digit machine: eight-digit accuracy versus six-digit accuracy.

- General Problem

- Find  $n$ -point difference approximation for  $f^{(k)}(x)$
- Optimal step size can be determined by Taylor-series expansions and linear equations.