

# Numerical Differentiation and Integration

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# Numerical Differentiation

- Forward differencing:

- $f'(x) \simeq \frac{f(x+h)-f(x)}{h}$

- $\frac{\partial f(x)}{\partial x_i} \simeq \frac{f(x_1, \dots, x_i + h_i, \dots, x_n) - f(x_1, \dots, x_i, \dots, x_n)}{h_i}$

- Backward differencing:

- $f'(x) \simeq \frac{f(x) - f(x-h)}{h}$

- $\frac{\partial f(x)}{\partial x_i} \simeq \frac{f(x_1, \dots, x_i, \dots, x_n) - f(x_1, \dots, x_i - h_i, \dots, x_n)}{h_i}$

- Note:

- In practice, one uses forward or backward differences depending on whether we care more about left or right derivative

- $h = \max(|x|, 1) \sqrt{\epsilon}$  where  $\epsilon$  is the machine precision (about  $10^{-15}$  in the Matlab)

# Numerical Differentiation

- Centered differencing:

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

- $\frac{\partial f(x)}{\partial x_i} \simeq \frac{f(x_1, \dots, x_i + h_i, \dots, x_n) - f(x_1, \dots, x_i - h_i, \dots, x_n)}{2h_i}$

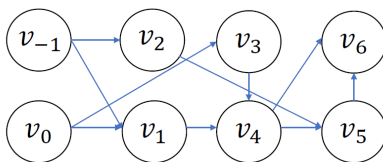
- Richardson's extrapolation (a fourth-order approximation):

$$f'(x) \simeq \frac{-f(x+2h) + 8f(x+h) - 8f(x-h) + f(x-2h)}{12h} \quad (1)$$

- Example:  $y = f(x_1, x_2) = [x_1^2 + x_1/x_2 - \exp(x_2)] [x_1/x_2 - \exp(x_2)]$  at  $(1, 1)$

# Automatic Differentiation

- Example:  $y = [x_1^2 + x_1/x_2 - \exp(x_2)] [x_1/x_2 - \exp(x_2)]$ 
  - Intermediate variables:  $v_{-1} = x_1$ ,  $v_0 = x_2$ ,  $v_1 = v_{-1}/v_0$ ,  $v_2 = v_{-1}^2$ ,  $v_3 = \exp(v_0)$ ,  
 $v_4 = v_1 - v_3$ ,  $v_5 = v_2 + v_4$ , and  $v_6 = v_4 \cdot v_5 = y$
  - Computational graph:



- By the chain rule

$$\begin{aligned}\frac{\partial y}{\partial x_1} &= \frac{\partial y}{\partial v_6} \left( \frac{\partial v_6}{\partial v_4} \frac{\partial v_4}{\partial v_1} \frac{\partial v_1}{\partial v_{-1}} + \frac{\partial v_6}{\partial v_5} \left( \frac{\partial v_5}{\partial v_4} \frac{\partial v_4}{\partial v_1} \frac{\partial v_1}{\partial v_{-1}} + \frac{\partial v_5}{\partial v_2} \frac{\partial v_2}{\partial v_{-1}} \right) \right) \frac{\partial v_{-1}}{\partial x_1} \\ \frac{\partial y}{\partial x_2} &= \frac{\partial y}{\partial v_6} \left( \frac{\partial v_6}{\partial v_4} \left( \frac{\partial v_4}{\partial v_1} \frac{\partial v_1}{\partial v_0} + \frac{\partial v_4}{\partial v_3} \frac{\partial v_3}{\partial v_0} \right) + \frac{\partial v_6}{\partial v_5} \frac{\partial v_5}{\partial v_4} \frac{\partial v_4}{\partial v_1} \frac{\partial v_1}{\partial v_0} \right) \frac{\partial v_0}{\partial x_2}\end{aligned}\tag{2}$$

# Numerical Quadrature Methods

- Integration:

$$S = \int_I f(x) dx \quad (3)$$

- Quadrature:

$$\int_I f(x) dx \simeq \sum_{i=1}^n w_i f(x_i) \quad (4)$$

- Quadrature methods differ only in how the quadrature weights  $w_i$  and the quadrature nodes  $x_i$  are chosen

# Newton-Cotes Methods

- Integration:

$$\int_a^b f(x) dx \quad (5)$$

- Trapezoid rule:

- $x_i = a + (i - 1) h$  where  $h = (b - a) / n$

- $\int_a^b f(x) dx \simeq \sum_{i=1}^n w_i f(x_i)$  where  $w_1 = w_n = h/2$  and  $w_i = h$ , otherwise

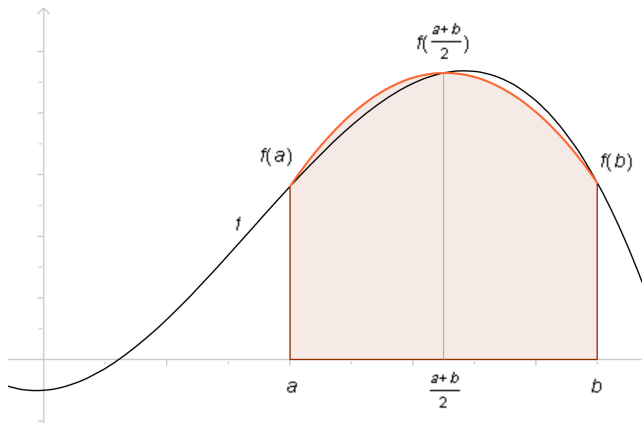
- Simpson's rule: piece-wise quadratic

- $x_i = a + (i - 1) h$  where  $h = (b - a) / (n - 1)$  and  $n$  is odd

- $\int_{x_{2j-1}}^{x_{2j+1}} f(x) dx \simeq \frac{h}{3} [f(x_{2j-1}) + 4f(x_{2j}) + f(x_{2j+1})]$

- $\int_a^b f(x) dx \simeq \sum_{i=1}^n w_i f(x_i)$  where  $w_1 = w_n = h/3$  and, otherwise,  $w_i = 4h/3$  if  $i$  is odd and  $w_i = 2h/3$  if  $i$  is even

# Simpson's rule



# Newton-Cotes Methods: Higher dimensional Integration

- Integration:

$$\int_{x_1 \in [a_1, b_1]} \int_{x_2 \in [a_2, b_2]} f(x_1, x_2) dx_1 dx_2 \quad (6)$$

- Newton-Cotes nodes and weights:

- $\{(x_{1i}, w_{1i}) \mid i = 1, 2, \dots, n_1\}$
- $\{(x_{2j}, w_{2j}) \mid j = 1, 2, \dots, n_2\}$
- Nodes:  $\{(x_{1i}, x_{2j}) \mid i = 1, 2, \dots, n_1; j = 1, 2, \dots, n_2\}$
- Weights:  $\{w_{ij} = w_{1i} w_{2j} \mid i = 1, 2, \dots, n_1; j = 1, 2, \dots, n_2\}$



# Gaussian Quadrature

- Integration:

$$\int_{-1}^1 f(x) dx = \sum_{i=1}^n w_i f(x_i) \quad (7)$$

- Nodes:  $x_i$  are the roots of a Legendre polynomial of degree  $n$ ,  $P_n(x)$

- Weights:  $w_i = \frac{2}{(1-x_i^2)[P'_n(x_i)]^2}$

- Legendre polynomials:

- Generating function:  $\frac{1}{\sqrt{1-2xt+t^2}} = \sum_{n=0}^{\infty} P_n(t)t^n$ , where  $P_0(x) = 1$  and  $P_1(x) = x$

- Recursive:  $(n+1)P_{n+1}(x) = (2n+1)xP_n(x) - nP_{n-1}(x)$

- Rodrigues' formula:  $P_n(x) = \frac{1}{2^n n!} \frac{d^n}{dx^n} (x^2 - 1)^n$

# Monte Carlo Integration

- Integration (multi-dimensional):  $S = \int_I f(x) dx$
- Let  $V = \int_I dx$
- $n$  uniform samples:  $x_1, \dots, x_n \in I$
- Then  $S \simeq \frac{V}{n} \sum_{i=1}^n f(x_i)$

## Numerical Integration: Example

- $H(x, y) = x^2 + y^2 + 2xy$
- Let  $I = [0, 1] \times [0, 2]$ . Compute

$$S = \int_I H(x, y) dx dy \quad (8)$$

- Analytical solution:  $S = \frac{16}{3}$
- Built-in integral function in Matlab: `quad2d`; `integral2`

## Example: Variable Markups in the Quantitative Trade Model

- $i = 1, \dots, N$  countries with labor  $(L_i)_{i=1}^N$
- The *Kimball's* preference over a continuum of varieties:

$$\int_{\omega \in \Omega_n} H\left(\frac{q_n(\omega)}{Q_n}\right) d\omega = 1, \quad (9)$$

where  $Q_n$  is the aggregate consumption and the function  $H(\cdot)$  is strictly increasing, strictly concave, and satisfies  $H(1) = 1$

- CES as a special case:  $H(q) = q^{\frac{\sigma-1}{\sigma}}$  for  $\sigma > 1$

## Example: Variable Markups in the Quantitative Trade Model

- The inverse demand function of variety  $\omega$  in country  $n$  can be expressed as

$$\frac{p_n(\omega)}{P_n} = H' \left( \frac{q_n(\omega)}{Q_n} \right) D_n, \quad (10)$$

where the *demand index*

$$D_n = \left[ \int_{\omega \in \Omega_n} H' \left( \frac{q_n(\omega)}{Q_n} \right) \frac{q_n(\omega)}{Q_n} d\omega \right]^{-1}, \quad (11)$$

and the price index for final goods

$$P_n = \int_{\omega \in \Omega_n} p_n(\omega) \frac{q_n(\omega)}{Q_n} d\omega = \int_{\omega \in \Omega_n} p_n(\omega) H'^{-1} \left( \frac{p_n(\omega)}{P_n} \frac{1}{D_n} \right) d\omega \quad (12)$$

## Example: Variable Markups in the Quantitative Trade Model

- Klenow and Willis (2016):

$$H(q) = 1 + (\sigma - 1) \exp\left(\frac{1}{\varepsilon}\right) \varepsilon^{\frac{\sigma}{\varepsilon}-1} \left[ \Gamma\left(\frac{\sigma}{\varepsilon}, \frac{1}{\varepsilon}\right) - \Gamma\left(\frac{\sigma}{\varepsilon}, \frac{q^{\frac{\varepsilon}{\sigma}}}{\varepsilon}\right) \right], \quad (13)$$

with  $\sigma > 1$  and  $\varepsilon \geq 0$  and where  $\Gamma(s, x)$  denotes the upper incomplete Gamma function

$$\Gamma(s, x) := \int_x^\infty t^{s-1} e^{-t} dt \quad (14)$$

- $\varepsilon = 0$ : the CES case  $H(q) = q^{\frac{\sigma-1}{\sigma}}$
- We have

$$H'(q) = \frac{\sigma - 1}{\sigma} \exp\left(\frac{1 - q^{\frac{\varepsilon}{\sigma}}}{\varepsilon}\right) \quad (15)$$

## Example: Variable Markups in the Quantitative Trade Model

- Each variety  $\omega$  is produced by a firm using labor under monopolistic competition
- Fixed marketing cost of firms to serve market  $n$ :  $F_n$  in units of  $n$ 's production labor
- Iceberg trade cost from country  $\ell$  to  $n$ :  $\tau_{\ell n} \geq 1$  with  $\tau_{\ell \ell} = 1$
- After paying a fixed entry cost  $f^e$  in terms of country  $i$ 's labor, firm  $\omega$  draws a productivity  $\varphi_i(\omega)$  from

$$\Pr(\varphi_i(\omega) \leq \varphi) = 1 - T_i \varphi^{-\theta}, \quad (16)$$

with support  $\varphi \geq T_i^{\frac{1}{\theta}}$

## Example: Variable Markups in the Quantitative Trade Model

- Conditional on firm  $\omega$  from country  $i$  serving country  $n$ , its effective cost  $c_{in}(\omega) = \frac{w_i \tau_{in}}{\varphi_i(\omega)}$  satisfies

$$\Pr(c_{in}(\omega) \leq c) = \bar{T}_{in}^\theta c^\theta, \quad c \leq \frac{1}{\bar{T}_{in}}, \quad \bar{T}_{in} \equiv T_i^{\frac{1}{\theta}} (w_i \tau_{in})^{-1} \quad (17)$$

- The operating profit of firm  $\omega$  from country  $i$  serving market  $n$  is

$$\tilde{\pi}_{in}(\omega) = \max_{q_{in}(\omega) \geq 0} \left[ H' \left( \frac{q_{in}(\omega)}{Q_n} \right) D_n P_n - c_{in}(\omega) \right] q_{in}(\omega). \quad (18)$$

- Let  $s_{in}(\omega) := \frac{q_{in}(\omega)}{Q_n}$  be the relative output. The optimal price can be expressed as a markup over the unit cost:

$$p_{in}(\omega) = \mu(s_{in}(\omega)) c_{in}(\omega), \quad \mu(s_{in}(\omega)) := \frac{\sigma s_{in}(\omega)^{-\frac{\varepsilon}{\sigma}}}{\sigma s_{in}(\omega)^{-\frac{\varepsilon}{\sigma}} - 1}. \quad (19)$$

- Let  $X_n := P_n Q_n$ . Then the operating profit can be expressed as

$$\tilde{\pi}_{in}(\omega) = \frac{s_{in}(\omega)^{\frac{\varepsilon}{\sigma}}}{\sigma} \frac{\sigma - 1}{\sigma} \exp \left( \frac{1 - s_{in}(\omega)^{\frac{\varepsilon}{\sigma}}}{\varepsilon} \right) s_{in}(\omega) D_n X_n. \quad (20)$$



## Example: Variable Markups in the Quantitative Trade Model

- Combining Equation (10) and (19), we can express  $s_{in}(\omega)$  in terms of  $c_{in}(\omega)$ :

$$\frac{\sigma}{\sigma - s_{in}(\omega)^{\frac{\varepsilon}{\sigma}}} c_{in}(\omega) = \frac{\sigma - 1}{\sigma} \exp\left(\frac{1 - s_{in}(\omega)^{\frac{\varepsilon}{\sigma}}}{\varepsilon}\right) D_n P_n \Rightarrow s_{in}(\omega) = s_n(c_{in}(\omega)) \quad (21)$$

- Then firm  $\omega$  from country  $i$  will serve destination market  $n$  if and only if

$$\frac{s_n(c_{in}(\omega))^{\frac{\varepsilon}{\sigma}}}{\sigma} \frac{\sigma - 1}{\sigma} \exp\left(\frac{1 - s_n(c_{in}(\omega))^{\frac{\varepsilon}{\sigma}}}{\varepsilon}\right) s_n(c_{in}(\omega)) D_n X_n \geq w_n F_n \quad (22)$$

- We assume that the cost cut-off  $c_n^*$  above which firms from country  $i$  will not serve market  $n$  satisfies:

$$c_n^* < \min\left\{\frac{1}{\bar{T}_{in}}, \frac{\sigma - 1}{\sigma} \exp\left(\frac{1}{\varepsilon}\right) D_n P_n\right\}, \quad \forall(i, n) \quad (23)$$

## Example: Variable Markups in the Quantitative Trade Model

- Let  $X_{in}$  be the total sales of firms originated from country  $i$  in destination market  $n$  and  $\Pi_{in}$  be the associated profit. Let  $M_i$  be the mass of firms in country  $i$

$$\lambda_{in} \equiv \frac{X_{in}}{X_n} = \frac{M_i \bar{T}_{in}^\theta}{\sum_{k=1}^N M_k \bar{T}_{kn}^\theta} \quad (24)$$

$$\frac{\Pi_{in}}{X_{in}} = \eta_n = \frac{\int_0^{c_n^*} \frac{s_n(c)^{\frac{\varepsilon}{\sigma}}}{\sigma} \frac{\sigma-1}{\sigma} \exp\left[\frac{1-s_n(c)^{\frac{\varepsilon}{\sigma}}}{\varepsilon}\right] s_n(c) c^{\theta-1} dc}{\int_0^{c_n^*} \frac{\sigma-1}{\sigma} \exp\left[\frac{1-s_n(c)^{\frac{\varepsilon}{\sigma}}}{\varepsilon}\right] s_n(c) c^{\theta-1} dc} \quad (25)$$

- The price index can be given by

$$P_n = \sum_{i=1}^N \nu_{in}^P M_i \bar{T}_{in}^\theta, \quad \nu_{in}^P \equiv \left[ \theta \int_0^{c_n^*} \frac{\sigma}{\sigma - s_n(c)^{\frac{\varepsilon}{\sigma}}} c s_n(c) c^{\theta-1} dc \right] \quad (26)$$

- By the definition of  $H(\cdot)$ , we have

$$\sum_{i=1}^N M_i \bar{T}_{in}^\theta \theta \int_0^{c_n^*} H(s(c)) c^{\theta-1} dc = 1 \quad (27)$$

## Example: Variable Markups in the Quantitative Trade Model

- Equilibrium consists of  $(w_i, M_i, P_i, D_i, c_i^*)_{i=1}^N$  such that

1.  $(w_i)$  is determined by labor market clearing

$$w_i L_i = \sum_{n=1}^N (1 - \eta_n) \lambda_{in} X_n + w_i F_i (c_i^*)^\theta \sum_{k=1}^N M_k \bar{T}_{ki}^\theta + \sum_{n=1}^N [\eta_n \lambda_{in} X_n - w_n F_n (c_n^*)^\theta M_i \bar{T}_{in}^\theta] \quad (28)$$

2. Firm mass  $M_i$  is determined by the free-entry condition:

$$M_i w_i f^e = \sum_{n=1}^N [\eta_n \lambda_{in} X_n - w_n F_n (c_n^*)^\theta M_i \bar{T}_{in}^\theta] \quad (29)$$

3. Total absorption:  $X_i = w_i L_i$
4. The price index is determined by Equation (26)
5.  $(c_n^*, D_n)$  are jointly determined by Equation (22) and (27)

## Example: Variable Markups in the Quantitative Trade Model

- Draw  $J$  numbers from the uniform distribution  $U[0, 1]$ , sorting them as  $u_1 < u_2 < \dots < u_J$
- Initial guess  $(w_i, M_i, P_i, D_i)_{i=1}^N$
- Compute the relative quantity cutoff below which firms will not serve the market  $n$  by solving:

$$\frac{(s_n^*)^{\frac{\varepsilon}{\sigma}}}{\sigma} \frac{\sigma - 1}{\sigma} \exp\left(\frac{1 - (s_n^*)^{\frac{\varepsilon}{\sigma}}}{\varepsilon}\right) s_n^* D_n X_n = w_n F_n \quad (30)$$

- The cost cutoff can then be computed by

$$c_n^* = \frac{\sigma - (s_n^*)^{\frac{\varepsilon}{\sigma}}}{\sigma} \frac{\sigma - 1}{\sigma} \exp\left(\frac{1 - (s_n^*)^{\frac{\varepsilon}{\sigma}}}{\varepsilon}\right) D_n P_n \quad (31)$$

- Compute  $\bar{T}_{in}$  by its definition and  $\lambda_{in}$  by Equation (24)

## Example: Variable Markups in the Quantitative Trade Model

- Let  $c_n^j = u_j c_n^*$ . The corresponding relative quantity  $s_n^j$  can be solved by Equation (21)

- Compute

$$\eta_n = \frac{\sum_{j=1}^J \frac{(s_n^j)^{\frac{\varepsilon}{\sigma}}}{\sigma} \exp \left[ \frac{1 - (s_n^j)^{\frac{\varepsilon}{\sigma}}}{\varepsilon} \right] s_n^j (c_n^j)^{\theta-1}}{\sum_{j=1}^J \exp \left[ \frac{1 - (s_n^j)^{\frac{\varepsilon}{\sigma}}}{\varepsilon} \right] s_n^j (c_n^j)^{\theta-1}} \quad (32)$$

- Compute

$$\nu_n^P = \frac{c_n^*}{J} \sum_{j=1}^J \frac{\theta \sigma}{\sigma - (s_n^j)^{\frac{\varepsilon}{\sigma}}} s_n^j (c_n^j)^{\theta} \quad (33)$$

- Update  $D_n$  by:

$$D_n = D_n \times \left[ \sum_{i=1}^N M_i \bar{T}_{in}^{\theta} \frac{c_n^*}{J} \theta \sum_{j=1}^J H(s_n^j) (c_n^j)^{\theta-1} \right]^{\frac{1}{1+\theta}} \quad (34)$$

## Example: Variable Markups in the Quantitative Trade Model

- $(w_i, M_i, P_i)$  are updated, respectively, by Equation (28), (29), and (26). Repeat until convergence. Notice that we update  $P_n$  by the following equation:

$$P_n^{(t+1)} = \left[ \left( P_n^{(t)} \right)^{-\theta-1} \nu_n^P \sum_{i=1}^N M_i \bar{T}_{in}^\theta \right]^{-\frac{1}{\theta}} \quad (35)$$

## Example: Variable Markups in the Quantitative Trade Model

- Welfare:  $U_i = \frac{w_i}{P_i}$
- Aggregate markup: as suggested by [Edmond et al. \(2019\)](#), we compute a sales-weighted harmonic average:

$$\begin{aligned}\bar{\mu}_n^D &:= \left[ \theta \left( \sum_{i=1}^N M_i \bar{T}_{in}^\theta \right) \int_0^{c_n^*} \left( \frac{\sigma}{\sigma - s_n(c)^{\frac{\varepsilon}{\sigma}}} \right)^{-1} \frac{p_n(c)}{P_n} s_n(c) c^{\theta-1} dc \right]^{-1} \\ &= \left[ \theta \left( \sum_{i=1}^N M_i \bar{T}_{in}^\theta \right) \frac{1}{P_n} \int_0^{c_n^*} s_n(c) c^\theta dc \right]^{-1} \\ &= \left[ \frac{\theta}{\nu_n^P} \int_0^{c_n^*} s_n(c) c^\theta dc \right]^{-1}\end{aligned}\tag{36}$$

## Example: Variable Markups in the Quantitative Trade Model

- Quantification:
  - How do changes in trade costs,  $\tau_{in}$ , affect markups in exporting country  $i$  as well as importing country  $n$ ?
  - How do welfare gains from trade rely on the markup variations? e.g. Comparing with the constant-markup models?



# Summary

- Numerical differentiation:
  - Crucial for nonlinear solvers and nonlinear optimization
  - Trade-off between accuracy and efficiency of computation
  - Promising direction: automatic differentiation and deep learning
- Numerical integration:
  - Heterogeneous-agent models
  - Random variables  $\Rightarrow$  Moments