

Numerical Methods in Economics

MIT Press, 1998

Chapter 13 Notes: Regular Perturbations of Simple Systems

November 15, 2006

Taylor Series

- Suppose that $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$.

- Linear approximation at $x = x_0$ is

$$f(x) \doteq f(x_0) + f_x(x_0)(x - x_0)$$

- Taylor series approximation of $f(x)$ based at x_0

$$f(x) \doteq f(x_0) + f_x(x_0)(x - x_0) + \frac{1}{2}(x - x_0)^\top f_{xx}(x_0)(x - x_0) + \dots$$

and computed via repeated differentiation.

- A function f and its Taylor series at x_0 have equal low-order derivatives at x_0 .
- Taylor series is a numerical approximation technique, useful when f and its derivatives are easily computed at some special point x_0 .

- Analytic Functions

- A function $f : \mathbb{R}^n \rightarrow \mathbb{R}$ is *analytic* at x iff f equals a power series on some open neighborhood of x .
- If analytic, $f(x)$ equals its infinite-term Taylor series expansion at x
- We approximate $f(x)$ for x near x_0 with first n terms of Taylor series expansion of f at x_0

Implicit Function Theorem

- Suppose $h : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is defined in $H(x, h(x)) = 0$, $H : \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}^m$, and $h(x_0) = y_0$.

– Implicit differentiation shows

$$H_x(x, h(x)) + H_y(x, h(x))h_x(x) = 0$$

– At $x = x_0$, this implies

$$h_x(x_0) = -H_y(x_0, y_0)^{-1}H_x(x_0, y_0)$$

if $H_y(x_0, y_0)$ is nonsingular. More simply, we express this as

$$h_x^0 = - (H_y^0)^{-1} H_x^0$$

– Linear approximation for $h(x)$ is

$$h^L(x) \doteq h(x_0) + h_x(x_0)(x - x_0)$$

- To check on quality, we compute

$$E = \hat{H}(x, h^L(x))$$

where \hat{H} is a unit free equivalent of H . If $E < \varepsilon$, then we have an ε -solution.

- If $h^L(y)$ is not satisfactory, compute higher-order terms by repeated differentiation.

– $D_{xx}H(x, h(x)) = 0$ implies

$$H_{xx} + 2H_{xy}h_x + H_{yy}h_x h_x + H_y h_{xx} = 0$$

– At $x = x_0$, this implies

$$h_{xx}^0 = - (H_y^0)^{-1} (H_{xx}^0 + 2H_{xy}^0 h_x^0 + H_{yy}^0 h_x^0 h_x^0)$$

– Construct the quadratic approximation

$$h^Q(x) \doteq h(x_0) + h_x^0(x - x_0) + \frac{1}{2}(x - x_0)^\top h_{xx}^0(x - x_0)$$

and check its quality by computing $E = H(x, h^Q(x))$.

Regular Perturbation: The Basic Idea

- Suppose x is an endogenous variable, ε a parameter

- Want to find $x(\varepsilon)$ such that $f(x(\varepsilon), \varepsilon) = 0$

- Suppose $x(0)$ known.

- Use Implicit Function Theorem

- Apply implicit differentiation:

$$f_x(x(\varepsilon), \varepsilon)x'(\varepsilon) + f_\varepsilon(x(\varepsilon), \varepsilon) = 0 \quad (13.1.5)$$

- At $\varepsilon = 0$, $x(0)$ is known and (13.1.5) is linear in $x'(0)$ with solution

$$x'(0) = -f_x(x(0), 0)^{-1}f_\varepsilon(x(0), 0)$$

- Well-defined only if $f_x \neq 0$, a condition which can be checked at $x = x(0)$.

- The linear approximation of $x(\varepsilon)$ for ε near zero is

$$x(\varepsilon) \doteq x^L(\varepsilon) \equiv x(0) - f_x(x(0), 0)^{-1}f_\varepsilon(x(0), 0)\varepsilon \quad (13.1.6)$$

- Can continue for higher-order derivatives of $x(\varepsilon)$.

– Differentiate (13.1.5) w.r.t. ε

$$f_x x'' + f_{xx} (x')^2 + 2f_{x\varepsilon} x' + f_{\varepsilon\varepsilon} = 0 \quad (13.1.7)$$

– At $\varepsilon = 0$, (13.1.7) implies that

$$\begin{aligned} x''(0) = & -f_x(x(0), 0)^{-1} \left(f_{xx}(x(0), 0) (x'(0))^2 \right. \\ & \left. + 2f_{x\varepsilon}(x(0), 0) x'(0) + f_{\varepsilon\varepsilon}(x(0), 0) \right) \end{aligned}$$

– Quadratic approximation is

$$x(\varepsilon) \doteq x^Q(\varepsilon) \equiv x(0) + \varepsilon x'(0) + \frac{1}{2} \varepsilon^2 x''(0) \quad (13.1.8)$$

- General Perturbation Strategy

- Find special (likely degenerate, uninteresting) case where one knows solution
 - * General relativity theory: begin with case of a universe with zero mass: ε is mass of universe
 - * Quantum mechanics: begin with case where electrons do not repel each other: ε is force of repulsion
 - * Business cycle analysis: begin with case where there are no shocks: ε is measure of exogenous shocks
- Use local approximation theory to compute nearby cases
 - * Standard implicit function may be applicable
 - * Sometimes standard implicit function theorem will not apply; use appropriate bifurcation or singularity method.
- Check to see if solution is good for problem of interest
 - * Use unit-free formulation of problem
 - * Go to higher-order terms until error is reduced to acceptable level
 - * *Always* check solution for range of validity

Single-Sector, Deterministic Growth - canonical problem

Consider dynamic programming problem

$$\max_{c(t)} \int_0^{\infty} e^{-\rho t} u(c) dt$$
$$\dot{k} = f(k) - c$$

Ad-Hoc Method: Convert to an LQ problem:

- Popular idea in macroeconomics
 - Compute steady state of deterministic problem
 - Convert nonlinear deterministic problem to LQ problem around steady state
 - Apply LQ methods to the approximate problem
 - Add noise to law of motion to get stochastic approximation

- McGrattan, JBES (1990)

- Replace $u(c)$ and $f(k)$ with approximations around c^* and k^*
- Solve linear-quadratic problem

$$\begin{aligned} \max_c \int_0^\infty e^{-\rho t} \left(u(c^*) + u'(c^*)(c - c^*) + \frac{1}{2}u''(c^*)(c - c^*)^2 \right) dt \\ \text{s.t. } \dot{k} = f(k^*) + f'(k^*)(k^* - k) - c \end{aligned}$$

- Resulting approximate policy function is

$$C^{McG}(k) = f(k^*) + \rho(k - k^*) \neq C(k^*) + C'(k^*)(k - k^*)$$

- * Local approximate law of motion is $\dot{k} = 0$
- * Add noise to get

$$dk = 0 \cdot dt + dz$$

- * Approximation is *random walk* when theory says solution is stationary

- Recent papers by Benigno and Woodford repeat this point.

- Kydland-Prescott

- Restate problem so that \dot{k} is linear function of state and controls
- Replace $u(c)$ with quadratic approximation
- Note 1: such transformation may not be easy
- Note 2: special case of Magill (JET 1977).

- Lesson

- Kydland-Prescott, McGrattan provide no mathematical basis for method
- Formal calculations based on appropriate IFT should be used.
- Beware of *ad hoc* methods based on an intuitive story!

Perturbation Method for Dynamic Systems

- Formalize problem as a system of functional equations

– Bellman equation:

$$\rho V(k) = \max_c u(c) + V'(k)(f(k) - c) \quad (1)$$

– $C(k)$: policy function defined by

$$\begin{aligned} 0 &= u'(C(k)) - V'(k) \\ \rho V(k) &= u(C(k)) + V'(k)(f(k) - C(k)) \end{aligned} \quad (2)$$

– Apply envelope theorem to (1) to get

$$\rho V'(k) = V''(k)(f(k) - C(k)) + V'(k)f'(k) \quad (1_k)$$

– Steady-state equations

$$\begin{aligned} c^* &= f(k^*) \\ 0 &= u'(c^*) - V'(k^*) \\ \rho V(k^*) &= u(c^*) + V'(k^*)(f(k^*) - c^*) \\ \rho V'(k) &= V''(k)(f(k^*) - c^*) + V'(k)f'(k) \end{aligned}$$

– Steady State: We know k^* , $V(k^*)$, $C(k^*)$, $f'(k^*)$, $V'(k^*)$:

$$\rho = f'(k^*)$$

$$C(k^*) = f(k^*)$$

$$V(k^*) = \rho^{-1}u(c^*)$$

$$V'(k^*) = u'(c^*)$$

– Want Taylor expansion:

$$C(k) \doteq C(k^*) + C'(k^*)(k - k^*) + C''(k^*)(k - k^*)^2/2 + \dots$$

$$V(k) \doteq V(k^*) + V'(k^*)(k - k^*) + V''(k^*)(k - k^*)^2/2 + \dots$$

- Linear approximation around a steady state

- Differentiate (1_k, 2) w.r.t. k :

$$\rho V'' = V'''(f - C) + V''(f' - C') + V''f' + V'f'' \quad (1_{kk})$$

$$0 = u''C' - V'' \quad (2_k)$$

- At the steady state

$$0 = -V''(k^*)C'(k^*) + V''(k^*)f'(k^*) + V'(k^*)f''(k^*) \quad (1_k^*)$$

- Substituting (2_k) into (1_k^{*}) yields

$$0 = -u''(C')^2 + u''C'f' + V'f''$$

- Two solutions

$$C'(k^*) = \frac{\rho}{2} \left(1 \pm \sqrt{1 + \frac{4u'(C(k^*))f''(k^*)}{u''(C(k^*))f'(k^*)f'(k^*)}} \right)$$

- However, we know $C'(k^*) > 0$; hence, take positive solution

- Higher-Order Expansions

- Conventional perception in macroeconomics: “perturbation methods of order higher than one are considerably more complicated than the traditional linear-quadratic case ...” – Marcat (1994, p. 111)
- Mathematics literature: No problem (See, e.g., Bensoussan, Fleming, Souganides, etc.)

- Compute $C''(k^*)$ and $V'''(k^*)$.

- Differentiate $(1_{kk}, 2_k)$:

$$\rho V''' = V''''(f - C) + 2V'''(f' - C') + V''(f'' - C'') \quad (1_{kkk})$$

$$+ V'''f' + 2V''f'' + V'f'''$$

$$0 = u'''(C')^2 + u''C'' - V''' \quad (2_{kk})$$

- At k^* , (1_{kkk}) reduces to

$$0 = 2V'''(f' - C') + 3V''f'' - V''C'' + V'f''' \quad (1_{kkk}^*)$$

- Equations $(1_{kkk}^*, 2_{kk}^*)$ are *LINEAR* in unknowns $C''(k^*)$ and $V'''(k^*)$:

$$\begin{pmatrix} u'' & -1 \\ V'' - 2(f' - C') \end{pmatrix} \begin{pmatrix} C'' \\ V''' \end{pmatrix} = \begin{pmatrix} A_1 \\ A_2 \end{pmatrix}$$

- Unique solution since determinant $-2u''(f' - C') + V'' < 0$.

- Compute $C^{(n)}(k^*)$ and $V^{(n+1)}(k^*)$.

- Linear system for order n is, for some A_1 and A_2 ,

$$\begin{pmatrix} u'' & -1 \\ V'' - n(f' - C') \end{pmatrix} \begin{pmatrix} C^{(n)} \\ V^{(n+1)} \end{pmatrix} = \begin{pmatrix} A_1 \\ A_2 \end{pmatrix}$$

- Higher-order terms are produced by solving linear systems

- The linear system is always determinate since $-nu''(f' - C') + V'' < 0$

- Conclusion:

- Computing first-order terms involves solving quadratic equations

- Computing higher-order terms involves solving linear equations

- Computing higher-order terms is easier than computing the linear term.

Accuracy Measure

Consider the one-period relative Euler equation error:

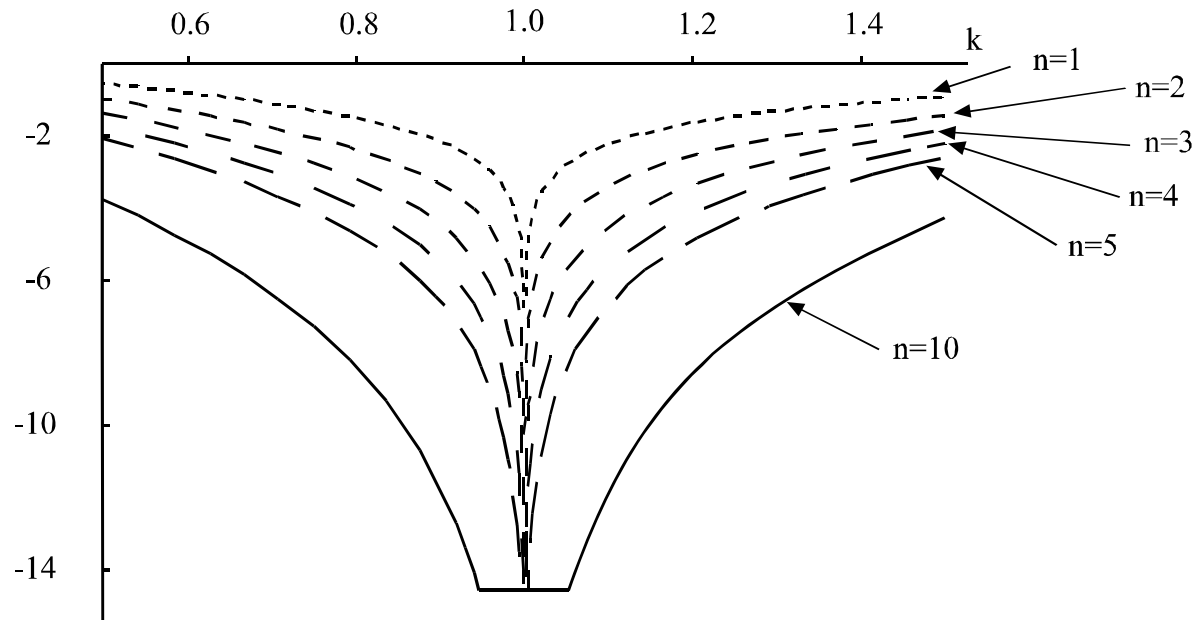
$$E(k) = 1 - \frac{V'(k)}{u'(C(k))}$$

- Equilibrium requires it to be zero.
- $E(k)$ is measure of optimization error
 - 1 is unacceptably large
 - Values such as .00001 is a limit for people.
 - $E(k)$ is unit-free.
- Define the L^p , $1 \leq p < \infty$, *bounded rationality accuracy* to be

$$\log_{10} \| E(k) \|_p$$

- The L^∞ error is the maximum value of $E(k)$.

Global Quality of Asymptotic Approximations



Graph of $\log_{10} |E(k)|$

- Linear approximation is very poor even for k close to steady state
- Order 2 is better but still not acceptable for even $k = .9, 1.1$
- Order 10 is excellent for $k \in [.6, 1.4]$