

Variational Methods & Optimal Control

lecture 12

Matthew Roughan

<matthew.roughan@adelaide.edu.au>

Discipline of Applied Mathematics
School of Mathematical Sciences
University of Adelaide

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Euler's finite difference method

We can approximate our function (and hence the integral) onto a finite grid. In this case, the problem reduces to a standard multivariable maximization (or minimization) problem, and we find the solution by setting the derivatives to zero. In the limit as the grid gets finer, this approximates the E-L equations.

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Numerical Solutions

The E-L equations may be hard to solve

Natural response is to find numerical methods

- ▶ Numerical solution of E-L DE
 - ▷ we won't consider these here (see other courses)
- ▶ Euler's finite difference method
- ▶ Ritz (Rayleigh-Ritz)
 - ▷ In 2D: Kantorovich's method

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Numerical Approximation

Numerical approximation of integrals:

- ▶ use an arbitrary set of mesh points $a = x_0 < x_1 < x_2 < \dots < x_n = b$.

- ▶ approximate

$$y'(x_i) = \frac{y_{i+1} - y_i}{x_{i+1} - x_i} = \frac{\Delta y_i}{\Delta x_i}$$

- ▶ rectangle rule

$$F\{y\} = \int_a^b f(x, y, y') dx \simeq \sum_{i=0}^{n-1} f\left(x_i, y_i, \frac{\Delta y_i}{\Delta x_i}\right) \Delta x_i = \bar{F}(y)$$

$\bar{F}(\cdot)$ is a function of the vector $y = (y_1, y_2, \dots, y_n)$.

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Finite Difference Method (FDM)

Treat this as a maximization of a function of n variables, so that we require

$$\frac{\partial \bar{F}}{\partial y_i} = 0$$

for all $i = 1, 2, \dots, n$.

Typically use uniform grid so $\Delta x_i = \Delta x = (b - a)/n$.

Simple Example

Find extremals for

$$F\{y\} = \int_0^1 \left[\frac{1}{2}y'^2 + \frac{1}{2}y^2 - y \right] dx$$

with $y(0) = 0$ and $y(1) = 0$.

E-L equations $y'' - y = 1$.

Simple Example: direct solution

E-L equations $y'' - y = -1$

Solution to homogeneous equations $y'' - y = 0$ is given by $e^{\lambda x}$ giving characteristic equation $\lambda^2 - 1 = 0$, so $\lambda = \pm 1$.

Particular solution $y = 1$

Final solution is

$$y(x) = Ae^x + Be^{-x} + 1$$

The boundary conditions $y(0) = y(1) = 0$ constrain $A + B = -1$ and $Ae + Be^{-1} = -1$, so $Ae + (1 - A)e^{-1} = 1$, so $A = \frac{e^{-1} - 1}{e - e^{-1}}$ and $B = \frac{1 - e}{e - e^{-1}}$.

Then the exact solution to the extremal problem is

$$y(x) = \frac{e^{-1} - 1}{e - e^{-1}} e^x + \frac{1 - e}{e - e^{-1}} e^{-x} - 1$$

Simple Example: Euler's FDM

Find extremals for

$$F\{y\} = \int_0^1 \left[\frac{1}{2}y'^2 + \frac{1}{2}y^2 - y \right] dx$$

Euler's FDM.

- ▶ Take the grid $x_i = i/n$, for $i = 0, 1, \dots, n$ so
 - ▷ end points $y_0 = 0$ and $y_n = 0$
 - ▷ $\Delta x = 1/n$
 - ▷ $\Delta y_i = y_{i+1} - y_i$
- ▶ So
 - ▷ $y'_i = \Delta y_i / \Delta x = n(y_{i+1} - y_i)$
 - ▷ and

$$y'_i{}^2 = n^2 (y_i^2 - 2y_i y_{i+1} + y_{i+1}^2)$$

Simple Example: Euler's FDM

Find extremals for

$$F\{y\} = \int_0^1 \left[\frac{1}{2}y'^2 + \frac{1}{2}y^2 - y \right] dx$$

Its FDM approximation is

$$\begin{aligned} \bar{F}(\mathbf{y}) &= \sum_{i=0}^{n-1} f(x_i, y_i, y'_i) \Delta x \\ &= \sum_{i=0}^{n-1} \frac{1}{2} n^2 (y_i^2 - 2y_i y_{i+1} + y_{i+1}^2) \Delta x + (y_i^2/2 - y_i) \Delta x \\ &= \sum_{i=0}^{n-1} \frac{1}{2} n (y_i^2 - 2y_i y_{i+1} + y_{i+1}^2) + \frac{y_i^2/2 - y_i}{n} \end{aligned}$$

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Simple Example: end-conditions

- ▶ We know the end conditions $y(0) = y(1) = 0$, which imply that

$$y_0 = y_n = 0$$

- ▶ Include them into the objective using Lagrange multipliers

$$\bar{H}(\mathbf{y}) = \sum_{i=0}^{n-1} \frac{1}{2} n (y_i^2 - 2y_i y_{i+1} + y_{i+1}^2) + \frac{y_i^2/2 - y_i}{n} + \lambda_0 y_0 + \lambda_n y_n$$

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Simple Example: Euler's FDM

Taking derivatives, note that y_i only appears in two terms of the FDM approximation

$$\begin{aligned} \bar{H}(\mathbf{y}) &= \sum_{i=0}^{n-1} \frac{1}{2} n (y_i^2 - 2y_i y_{i+1} + y_{i+1}^2) + \frac{y_i^2/2 - y_i}{n} + \lambda_0 y_0 + \lambda_n y_n \\ \frac{\partial \bar{H}(\mathbf{y})}{\partial y_i} &= \begin{cases} n(y_0 - y_1) + \frac{y_0 - 1}{n} + \lambda_0 & \text{for } i = 0 \\ n(2y_i - y_{i+1} - y_{i-1}) + \frac{y_i}{n} - \frac{1}{n} & \text{for } i = 1, \dots, n-1 \\ n(y_n - y_{n-1}) + \lambda_n & \text{for } i = n \end{cases} \end{aligned}$$

We need to set the derivatives to all be zero, so we now have $n+3$ linear equations, including $y_0 = y_n = 0$, and $n+3$ variables including the two Lagrange multipliers. We can solve this system numerically using, e.g., matlab.

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Simple Example: Euler's FDM

Example: $n = 4$, solve

$$A\mathbf{z} = \mathbf{b}$$

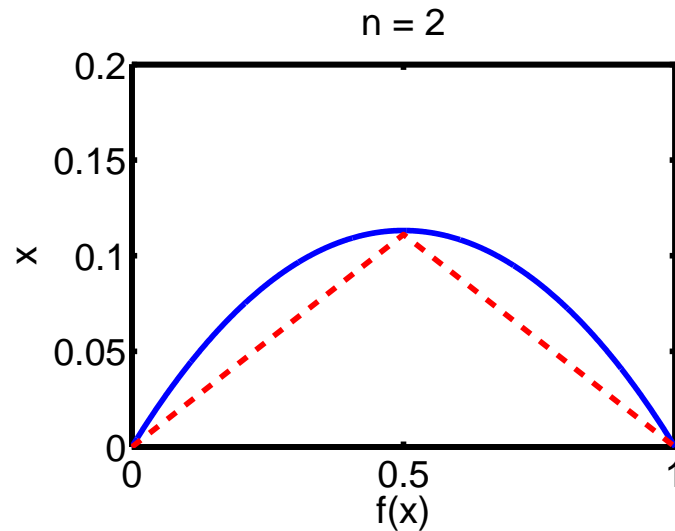
where

$$A = \begin{pmatrix} -4.00 & & & & & & & & & & \\ & 8.25 & -4.00 & & & & & & & & -4.00 \\ -4.00 & & 8.25 & -4.00 & & & & & & & \\ & & -4.00 & & 8.25 & -4.00 & & & & & \\ & & & -4.00 & & 8.25 & -4.00 & & & & \\ & & & & -4.00 & & 8.25 & & & & -4.00 \\ & & & & & & & -4.00 & & & \\ & & & & & & & & -4.00 & & \\ & & & & & & & & & -4.00 & \end{pmatrix} \quad \text{and } \mathbf{b} = \begin{pmatrix} 0.00 \\ 0.25 \\ 0.25 \\ 0.25 \\ 0.25 \\ 0.00 \\ 0.00 \end{pmatrix}$$

- ▶ first $n+1$ terms of \mathbf{z} give \mathbf{y}
- ▶ last two terms given the Lagrange multipliers λ_0 and λ_n

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Simple example: results



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Convergence of Euler's FDM

The condition for a stationary point becomes

$$\frac{\partial \bar{F}}{\partial y_i} = \frac{\partial f}{\partial y_i} (x_i, y_i, y_i') - \frac{\frac{\partial f}{\partial y_i'} (x_i, y_i, \frac{\Delta y_i}{\Delta x}) - \frac{\partial f}{\partial y_i'} (x_{i-1}, y_{i-1}, \frac{\Delta y_{i-1}}{\Delta x})}{\Delta x} = 0$$

In limit $n \rightarrow \infty$, then $\Delta x \rightarrow 0$, and so we get

$$\frac{\partial f}{\partial y} - \frac{d}{dx} \left(\frac{\partial f}{\partial y'} \right) = 0$$

which are the Euler-Lagrange equations.

- ▶ i.e., the finite difference solution converges to the solution of the E-L equations

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Convergence of Euler's FDM

$$\bar{F}(\mathbf{y}) = \sum_{i=0}^{n-1} f \left(x_i, y_i, \frac{\Delta y_i}{\Delta x} \right) \Delta x \quad \text{and} \quad \Delta y_i = y_{i+1} - y_i$$

Only and two terms in the sum involve y_i , so

$$\begin{aligned} \frac{\partial \bar{F}}{\partial y_i} &= \frac{\partial}{\partial y_i} f \left(x_{i-1}, y_{i-1}, \frac{\Delta y_{i-1}}{\Delta x} \right) + \frac{\partial}{\partial y_i} f \left(x_i, y_i, \frac{\Delta y_i}{\Delta x} \right) \\ &= \frac{1}{\Delta x} \frac{\partial f}{\partial y_i'} \left(x_{i-1}, y_{i-1}, \frac{\Delta y_{i-1}}{\Delta x} \right) \\ &\quad + \frac{\partial f}{\partial y_i} \left(x_i, y_i, \frac{\Delta y_i}{\Delta x} \right) - \frac{1}{\Delta x} \frac{\partial f}{\partial y_i'} \left(x_i, y_i, \frac{\Delta y_i}{\Delta x} \right) \\ &= \frac{\partial f}{\partial y_i} (x_i, y_i, y_i') - \frac{\frac{\partial f}{\partial y_i'} (x_i, y_i, \frac{\Delta y_i}{\Delta x}) - \frac{\partial f}{\partial y_i'} (x_{i-1}, y_{i-1}, \frac{\Delta y_{i-1}}{\Delta x})}{\Delta x} \end{aligned}$$

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Comments

- ▶ There are lots of ways to improve Euler's FDM
 - ▷ use a better method of numerical quadrature (integration)
 - ★ trapezoidal rule
 - ★ Simpson's rule
 - ★ Romberg's method
 - ▷ use a non-uniform grid
 - ★ make it finer where there is more variation
- ▶ We can use a different approach that can be even better

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Ritz's method

In Ritz's method (called Kantorovich's methods where there is more than one independent variable), we approximate our functions (the extremal in particular) using a family of simple functions. Again we can reduce the problem into a standard multivariable maximization problem, but now we seek coefficients for our approximation.

Ritz's method

- ▶ select $\{\phi_j\}_{j=0}^n$
- ▶ Approximate $y_n(x) = \phi_0(x) + c_1\phi_1(x) + c_2\phi_2(x) + \dots + c_n\phi_n(x)$
- ▶ Approximate $F\{y\} \simeq F\{y_n\} = \int_{x_0}^{x_1} f(x, y_n, y_n') dx$
- ▶ Integrate to get $F\{y_n\} = F_n(c_1, c_2, \dots, c_n)$
- ▶ F_n is a known function of n variables, so we can maximize (or minimize) it as usual by

$$\frac{\partial F_n}{\partial c_i} = 0$$

for all $i = 1, 2, \dots, n$.

Ritz's method

Assume we can approximate $y(x)$ by

$$y(x) = \phi_0(x) + c_1\phi_1(x) + c_2\phi_2(x) + \dots + c_n\phi_n(x)$$

where we choose a convenient set of functions $\phi_j(x)$ and find the values of c_j which produce an extremal.

For fixed end-point problem:

- ▶ Choose $\phi_0(x)$ to satisfy the end conditions.
- ▶ Then $\phi_j(x_0) = \phi_j(x_1) = 0$ for $j = 1, 2, \dots, n$

The ϕ can be chosen from standard sets of functions, e.g. power series, trigonometric functions, Bessel functions, etc. (but must be linearly independent)

Upper bounds

Assume the extremal of interest is a minimum, then for the extremal

$$F\{y\} < F\{\hat{y}\}$$

for all \hat{y} within the neighborhood of y . Assume our approximating function y_n is close enough to be in that neighborhood, then

$$F\{y\} \leq F\{y_n\} = F_n(\mathbf{c})$$

so the approximation provides an **upper bound** on the minimum $F\{y\}$. Another way to think about it is that we optimize on a smaller set of possible functions y , so we can't get quite as good a minimum.

Simple Example

Find extremals for

$$F\{y\} = \int_0^1 \left[\frac{1}{2}y'^2 + \frac{1}{2}y^2 - y \right] dx$$

with $y(0) = 0$ and $y(1) = 0$.

E-L equations $y'' - y = 1$, but we shall bypass the E-L equations to use Ritz's method.

$$y_n(x) = \phi_0(x) + \sum_{i=1}^n c_i \phi_i(x)$$

where we take $\phi_0(x) = 0$ and $\phi_i(x) = x^i(1-x)^i$.

Simple Example

We solve for c_1 by setting

$$\frac{dF_1}{dc_1} = \frac{11c_1}{30} - \frac{1}{6} = 0$$

to get $c_1 = 5/11$, so the approximate extremal is

$$y_1(x) = \frac{5}{11}x(1-x)$$

The value of the approximate functional at this point is

$$F_1(5/11) = \frac{c_1^2}{2} \frac{11}{30} - \frac{c_1}{6} = -0.37879$$

which is an upper bound on the true value of the functional on the extremal.

Simple Example

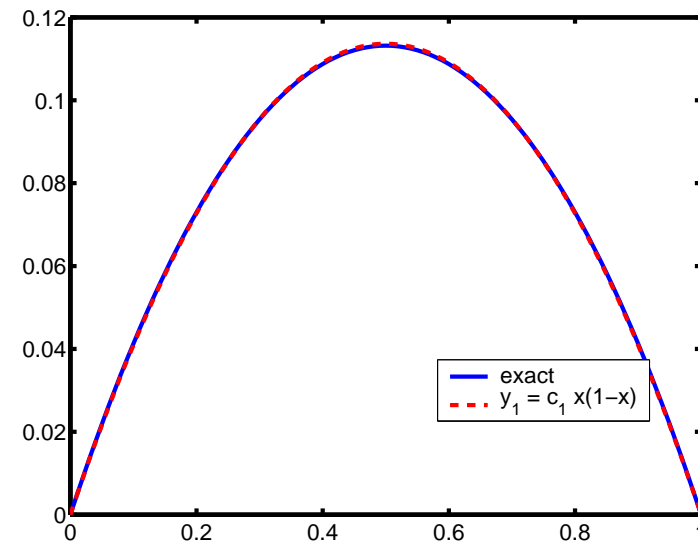
Simple approximation $y_1 = c_1 \phi_1(x)$ we get

$$F_1(c_1) = F\{y_1\} = \int_0^1 \left[\frac{1}{2}c_1^2 \phi_1'^2 + c_1^2 \frac{1}{2} \phi_1^2 - c_1 \phi_1 \right] dx$$

Now $\phi(x) = x(1-x)$ so $\phi_1' = 1 - 2x$, and

$$\begin{aligned} F_1(c_1) &= \int_0^1 \left[\frac{c_1^2}{2} (1-2x)^2 + \frac{c_1^2}{2} x^2(1-x)^2 - c_1 x(1-x) \right] dx \\ &= \frac{c_1^2}{2} \int_0^1 [1 - 4x + 5x^2 - x^4] dx + c_1 \int_0^1 [-x + x^2] dx \\ &= \frac{c_1^2}{2} [x - 2x^2 + 5x^3/3 - x^5/5]_0^1 + c_1 [-x^2/2 + x^3/3]_0^1 \\ &= \frac{c_1^2}{2} \frac{11}{30} - \frac{c_1}{6} \end{aligned}$$

Simple example: results



Alternate approach

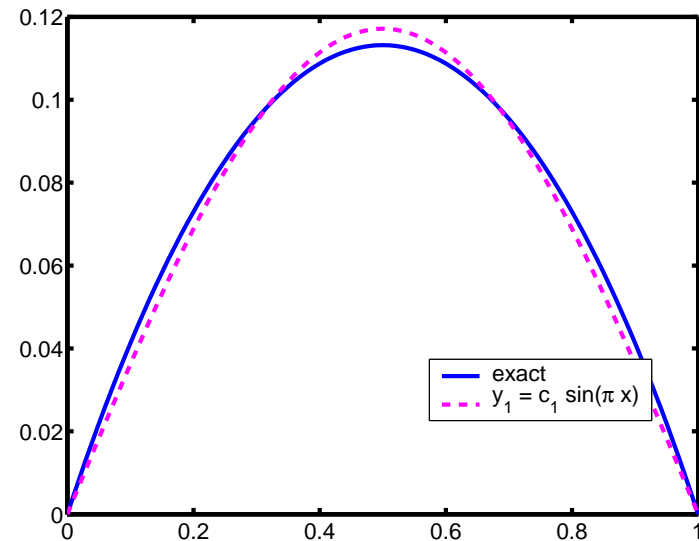
Choose $\phi_1(x) = \sin(\pi x)$ (use the first element of a trigonometric series to approximate y). Then, $\phi'(x) = \pi \cos(\pi x)$, and so the functional is

$$\begin{aligned} F_1(c_1) &= F\{c_1\phi_1\} = \int_0^1 \left[\frac{1}{2}c_1^2\phi_1'^2 + c_1^2\frac{1}{2}\phi_1^2 - c_1\phi_1 \right] dx \\ &= \int_0^1 \left[\frac{c_1^2\pi^2}{2}\cos^2(\pi x) + \frac{c_1^2}{2}\sin^2(\pi x) - c_1\sin(\pi x) \right] dx \end{aligned}$$

Now $\int_0^1 \cos^2(\pi x) = \int_0^1 \sin^2(\pi x) = 1/2$,
and $\int_0^1 \sin(\pi x) = [-\frac{1}{\pi}\cos(\pi x)]_0^1 = -2/\pi$, so

$$F(c_1) = \frac{c_1^2}{2} \frac{1}{2} [\pi^2 + 1] - \frac{2}{\pi}c_1$$

example: alternative results



Alternate approach

Once again we solve for c_1 by setting

$$\frac{dF_1}{dc_1} = c_1 \frac{1}{2} [\pi^2 + 1] - \frac{2}{\pi} = 0$$

to get $c_1 = \frac{4}{\pi(\pi^2+1)}$, so the approximate extremal is

$$y_1(x) = \frac{4}{\pi(\pi^2+1)} \sin(\pi x)$$