

Notes

- ◆ Extra class this Friday 1-2pm
- ◆ Assignment 2 is out
- ◆ Error in last lecture: quasi-Newton methods based on the secant condition:
$$H\Delta x \approx g(x + \Delta x) - g(x)$$
 - Really just Taylor series applied to the gradient function g :
$$g(x + \Delta x) = g(x) + H\Delta x + O(\Delta x^2)$$

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Nonlinear Least Squares

- ◆ One particular nonlinear problem of interest:

$$\min_x \|b - g(x)\|_2^2$$

- ◆ Can we exploit the structure of this objective?
- ◆ Gradient: $-2J^T(b - g(x))$, $J_{ij} = \frac{\partial g_i}{\partial x_j}$
- ◆ Hessian: $H = 2J^T J - 2(b - g(x)) \cdot \frac{\partial^2 g}{\partial x^2}$

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Gauss-Newton

- ◆ Assuming g is close to linear, just keep the $J^T J$ term
- ◆ Automatically positive (semi-)definite: in fact, modified Newton solve is now a linear least-squares problem!

$$\min_{\Delta x} \|(b - g(x)) - J\Delta x\|_2^2$$

- ◆ Convergence may not be quite as fast as Newton, but more robust.

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Time Integration

- ◆ A core problem across many applications
 - Physics, robotics, finance, control, ...
- ◆ Typical statement: system of ordinary differential equations, first order
$$\frac{dy}{dt} = f(y, t)$$

Subject to initial conditions:

$$y(0) = y_0$$

- ◆ Higher order equations can be reduced to a system - though not always a good thing to do

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Forward Euler

- ◆ Derive from Taylor series:

$$y(t + \Delta t) = y(t) + \Delta t \frac{dy}{dt}(t) + O(\Delta t^2)$$

$$y_{n+1} = y_n + \Delta t f(y_n, t_n)$$

- ◆ **Truncation error** is second-order
- ◆ (Global) accuracy is first-order
 - Heuristic: to get to a fixed time T , need $T/\Delta t$ time steps

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Analyzing Forward Euler

- ◆ Euler proved convergence as Δt goes to 0
- ◆ In practice, need to pick $\Delta t > 0$
 - how close to zero do you need to go?
- ◆ The first and foremost tool is using a simple model problem that we can solve exactly

$$\frac{dy}{dt} = Ay$$

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Solving Linear ODE's

- ◆ First: what is the exact solution?
- ◆ For scalars: $\frac{dy}{dt} = Ay, \quad y(0) = y_0$
 $\Rightarrow y(t) = e^{At}y_0$
- ◆ This is in fact true for systems as well, though needs the “matrix exponential”
- ◆ More elementary approach: change basis to diagonalize A (or reduce to Jordan canonical form)

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The Test Equation

- ◆ We can thus simplify our model down to a single scalar equation, **the test equation**

$$\frac{dy}{dt} = \lambda y$$

- ◆ Relate this to a full nonlinear system by taking the eigenvalues of the Jacobian of f:

$$\left(\frac{\partial f}{\partial y}\right)_x = \lambda x$$

- But obviously nonlinearities may cause unexpected things to happen: full nonlinear analysis is generally problem-specific, and hard or impossible

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Solution of the Test Equation

- ◆ Split scalar into real and imaginary parts:

$$\lambda = a + b\sqrt{-1}$$

- ◆ For initial conditions $y_0=1$

$$y(t) = e^{\lambda t} = e^{at} \left(\cos bt + \sqrt{-1} \sin bt \right)$$

- ◆ The magnitude only depends on a
 - If $a < 0$: exponential decay (stable)
 - If $a > 0$: exponential increase (unstable)
 - If $a = 0$: borderline-stable

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