

Solving ODEs in Python

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(From MATLAB slides by James Osborne)



Aim and contents



- Aim: Learn techniques for the solution of systems of Ordinary Differential Equations
- Contents:
 - Analytical methods for simple ODEs
 - Reducing the order of ODEs
 - Numerical methods for first order ODEs
 - Half-day exercise
 - Python for solving initial value problems
 - Python for solving boundary value problems



First order ODEs?



- ODE Ordinary Differential Equation,
 - With respect to one variable, t or x etc.
- Order of ODE order of the highest derivative
- First order ODE: $\frac{\mathrm{d}y}{\mathrm{d}x} = f(x,y), \, y(0) = a.$
- Simple problems solve analytically
 - Separable solutions, integrating factors
- Highly non-linear problems or unknown integral, then solve numerically
 - Forward Euler method, Runge-Kutta method...
 - In-built scipy (or other) solvers





Analytical methods



Analytical techniques



$$\frac{\mathrm{d}y}{\mathrm{d}x} = \frac{f(x)}{g(y)}$$
, Separable solutions

$$\frac{\mathrm{d}y}{\mathrm{d}x} + f(x)y = g(x)$$
, Integrating factor

$$a\frac{\mathrm{d}^2y}{\mathrm{d}x^2} + b\frac{\mathrm{d}y}{\mathrm{d}x} + cy = 0$$
, Auxiliary equation





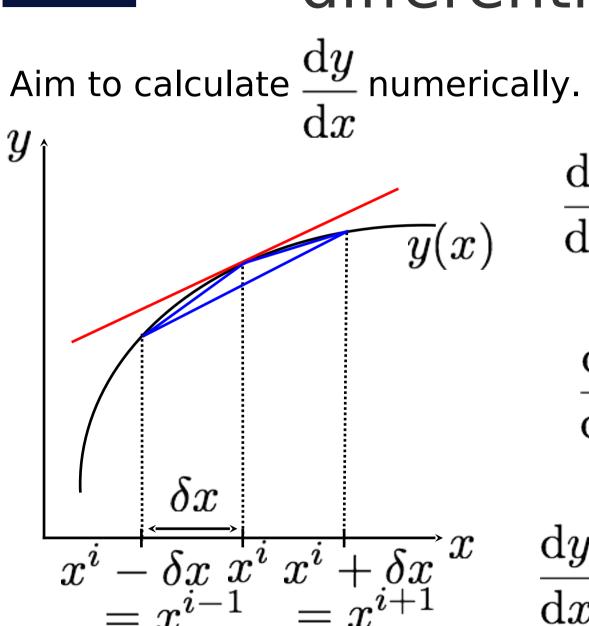
Numerical approaches



Numerical differentiation



Backward difference



$$\frac{\mathrm{d}y}{\mathrm{d}x} \approx \frac{y(x^i) - y(x^{i-1})}{\delta x},$$

Forward difference

$$\frac{\mathrm{d}y}{\mathrm{d}x} pprox \frac{y(x^{i+1}) - y(x^i)}{\delta x}$$

Central difference

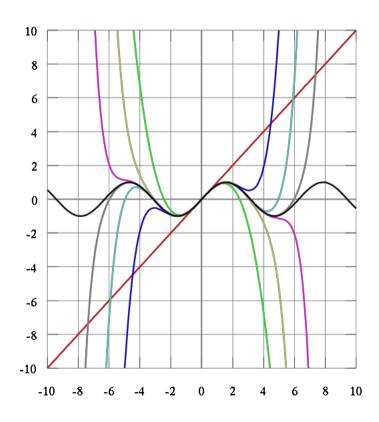
$$\frac{\mathrm{d}y}{\mathrm{d}x} pprox \frac{y(x^{i+1}) - y(x^{i-1})}{2\delta x}$$



Taylor expansion



$$f(x) = f(a) + (x - a)f'(a) + \frac{(x - a)^2 f''(a)}{2!} + \frac{(x - a)^3 f'''(a)}{3!} + \cdots$$



 $\sin(x)$

Use this to prove the finite difference formulas

From Wikipedia



Euler method

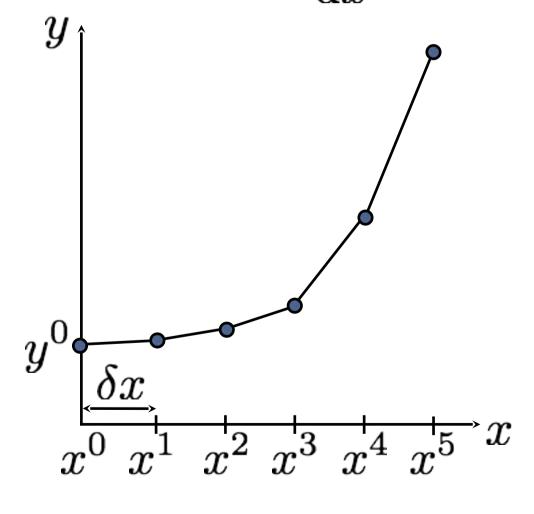


Want to solve

$$\frac{\mathrm{d}y}{\mathrm{d}x} = f(x, y),$$

such that

$$y(0) = a$$
.



$$x^{0} = 0,$$

 $x^{i} = x^{i-1} + \delta x$
 $y^{0} = a,$
 $y^{i} \approx y(x^{i}),$

$$\frac{y^{i+1} - y^i}{\delta x} = f(x, y).$$



Forward vs Backward Euler



$$\frac{y^{i+1} - y^i}{\delta x} = f(x^i, y^i),$$

Forward Euler method "Explicit"

$$(y^{i+1}) = y^i + \delta x f(x^i, y^i),$$

$$\frac{y^{i+1} - y^i}{\delta x} = f(x^{i+1}, y^{i+1}),$$

Backward Euler method "Implicit"

$$(y^{i+1}) - \delta x f(x^{i+1}(y^{i+1})) = y^i,$$

Forward - conditionally stable Backward - unconditionally stable



Euler method for systems of ODEs



Can extend this to systems of ODEs

$$\frac{dy_1}{dx} = f_1(x, y_1, y_2, y_3),$$

$$\frac{dy_2}{dx} = f_2(x, y_1, y_2, y_3),$$

$$\frac{dy_3}{dx} = f_3(x, y_1, y_2, y_3),$$

$$\frac{\mathrm{d}\mathbf{y}}{\mathrm{d}x} = \mathbf{f}(x, \mathbf{y}),$$

$$\mathbf{y}^i = (y_1^i, y_2^i, y_3^i), \qquad \frac{\mathrm{d}\mathbf{y}}{\mathrm{d}x} pprox \frac{\mathbf{y}^{i+1} - \mathbf{y}^i}{\delta x}.$$



Higher order ODEs



$$a(x)\frac{\mathrm{d}^2 y}{\mathrm{d}x^2} + b(x)\frac{\mathrm{d}y}{\mathrm{d}x} + c(x)y = f(x),$$

Reduce to a system of first order ODEs

$$\frac{\mathrm{d}y}{\mathrm{d}x} = z,$$
 System of first order
$$\frac{\mathrm{d}z}{\mathrm{d}x} = \frac{f(x) - b(x)z - c(x)y}{a(x)},$$
 ODEs



Improving on Euler



- Use error analysis to improve placement of nodes (adaptivity)
- Higher order methods:
 - Runge-Kutta;
 - Dormand-Prince;
 - Adams-Bashforth;
 - Adams-Moulton
- See Suli and Mayers "An Introduction to Numerical Analysis" for more details



Pause for "Plan"



- Write your own solver
 - Morning exercise
- Use Python to solve systems of ODEs
 - Afternoon...





Using Python



Python and ODEs: IVPs



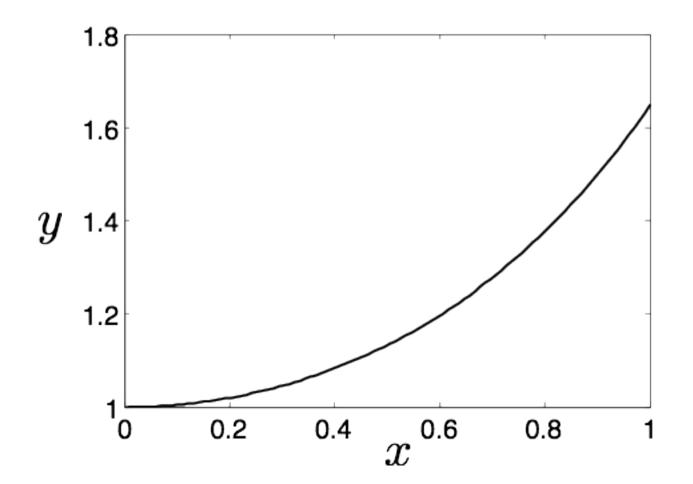
- Initial value problems
 - Using odeint or ode
 - odeint is easy to set up
 - ode is more configurable
 - Based on Runge-Kutta schemes
- Two examples:
 - Single ODE; and
 - Coupled system of ODEs



Example problem



$$\frac{\mathrm{d}y}{\mathrm{d}x} = xy, \ y(0) = 1, \longrightarrow y(x) = e^{\frac{x^2}{2}}.$$





odeint versus ode



```
# Function to solve dydx=x*y
def dydx1(y, x):
    # dydx=x*y
    return x*y

y0 = 1
xs = np.linspace(0, 1, 100)
ys = odeint(dydx1, y0, xs)

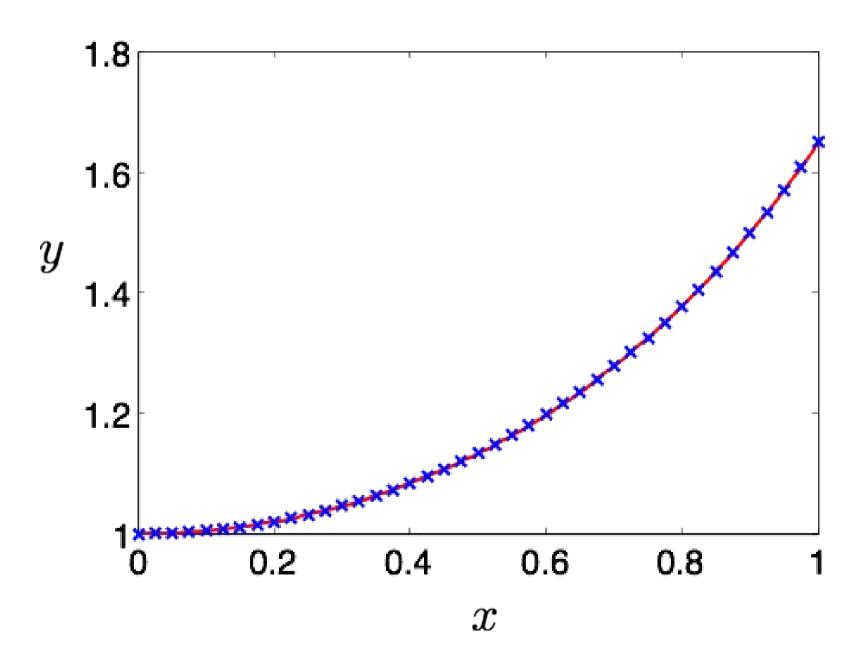
plt.plot(xs, ys)
plt.xlabel('x');
plt.ylabel('y')
plt.show()
```

```
# Function to solve dydx=x*y
def dydx2(x, y):
  # dydx=x*y
  return x*y
y0 = 1; x = 0
solver = ode(dydx2)
solver.set initial value(y0, x)
xs = [x]; ys = [y0]
while x<1:</pre>
    x += 0.01
    y=solver.integrate(x)
    ys.append(y[0])
    xs.append(x)
plt.plot(xs, ys)
plt.xlabel('x'); plt.ylabel('y')
plt.show()
```



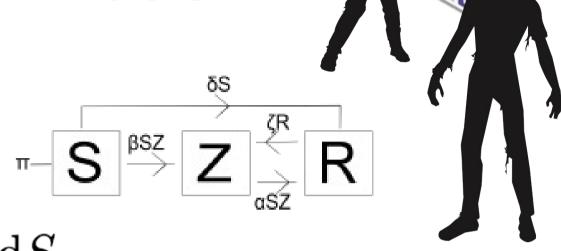
Results







Zombie model



 $\Pi - \beta SZ - \delta S$,

- "When zombies
 Attack!" (Munz et al.
 2009) presents model
 of zombie invasion
- System of 3 ODEs:
 - S Susceptibles
 - Z Zombies
 - R Removed

$$\frac{\mathrm{d}Z}{\mathrm{d}t} = \beta SZ + \zeta R - \alpha SZ,$$

$$\frac{\mathrm{d}R}{\mathrm{d}t} = \delta S + \alpha SZ - \zeta R.$$



Zombie code



```
# Function to solve the SZR Zombie ODE system.
alpha = 0.005; zeta = 0.1; pi = 0
beta = 0.0095: delta = 0.0001
def SZR(Y,t):
    S = Y[0]; Z = Y[1]; R = Y[2]
    #Susceptible / Zombie / Removed
    dSdt = pi - beta*S*Z - delta*S
    dZdt = beta*S*Z + zeta*R - alpha*S*Z
    dRdt = delta*S + alpha*S*Z - zeta*R
    return [dSdt, dZdt, dRdt]
S0=1000: Z0=0: R0=0
EndTime = 5
t = np.linspace(0, EndTime, 100)
Y0 = [S0, Z0, R0]
Y = odeint(SZR, Y0, t)
plt.plot(t,Y[:,0], 'b', label='Susceptible')
plt.plot(t,Y[:,1], 'r', label='Zombie')
plt.plot(t,Y[:,2], 'g', label='Removed')
plt.xlabel('Time');plt.ylabel('Population')
plt.legend();plt.show()
```

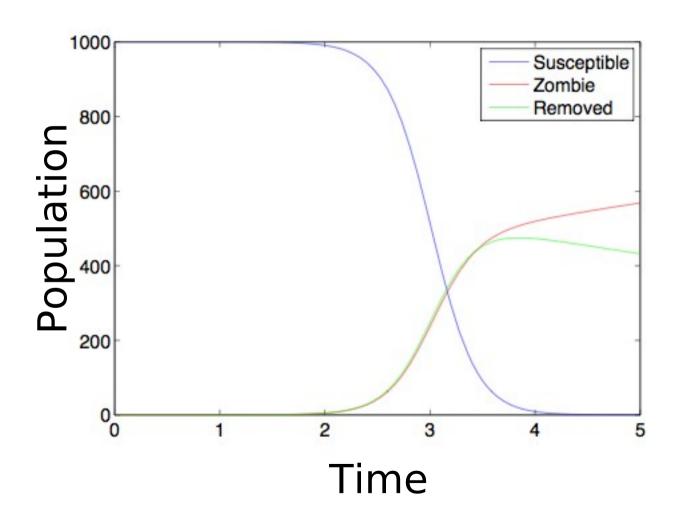


Zombie results



$$S_0 = 1000,$$

 $Z_0 = 0,$
 $R_0 = 0.$



More complicated models in exercises



Python and ODEs: BVPs



- Boundary value problems
 - The shooting method
- Simple example:

$$\frac{\mathrm{d}^2 y}{\mathrm{d} x^2} + y = 0$$
, with $y'(0) = 1, y(\pi) = 0$.

$$\left. egin{array}{l} rac{\mathrm{d} y}{\mathrm{d} x} = z, \\ rac{\mathrm{d} z}{\mathrm{d} x} = -y, \end{array}
ight\}$$
 First order system Exact solution $y = sin(x)$.



Simple BVP code

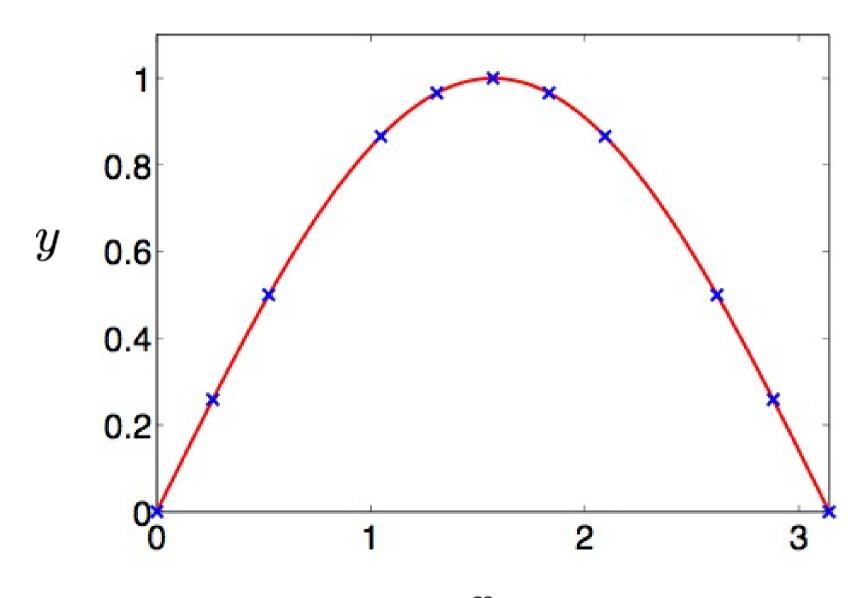


```
# Solving y'' + y = 0
# y[0] is y, y[1] is y'
\# dy[0]/dx = y[1] \text{ and } dy[1]/dx = -y[0]
def dydx(x, y):
     return np.vstack((y[1], -y[0]))
def bcs(yat0, yatpi):
    # Neumann/Dirichlet y'(x=0)=1, y(x=pi)=0
    return (yat0[1]-1, yatpi[0])
x = np.linspace(0, math.pi, 10)
init y = np.ones((2, x.size))
sol = solve_bvp(dydx, bcs, x, init_y)
plt.plot(sol.x, sol.y[0], 'b+')
xs = np.linspace(0, math.pi, 100)
plt.plot(xs, np.sin(xs), 'r')
plt.xlabel('x'); plt.ylabel('y')
plt.show()
```



Results







Overview



- Techniques for solution of simple ODEs
- Reduction of higher order systems to first order
- Numerical differentiation
- Use Python to solve systems of ODEs with
 - initial conditions, and
 - boundary conditions



Separable solutions



$$\frac{\mathrm{d}y}{\mathrm{d}x} = \frac{f(x)}{g(y)}, \qquad \int g(y)\mathrm{d}y = \int f(x)\mathrm{d}x + C,$$

Example

$$\frac{\mathrm{d}y}{\mathrm{d}x} = xy, \ y(0) = 1,$$

$$y(x) = e^{\frac{x^2}{2}}$$



Integrating factors



$$\frac{\mathrm{d}y}{\mathrm{d}x} + f(x)y = g(x), \qquad \phi(x) = e^{\{\int f(x)\mathrm{d}x\}},$$

Example

$$\frac{\mathrm{d}y}{\mathrm{d}x} + \frac{y}{x} = 1, \ x > 0, \ y(1) = 0,$$

$$\phi(x) = x, \quad y(x) = \frac{x}{2} - \frac{1}{2x}.$$



Second order ODE with linear coefficients



$$a\frac{\mathrm{d}^2 y}{\mathrm{d}x^2} + b\frac{\mathrm{d}y}{\mathrm{d}x} + cy = 0$$
, try $y(x) = Ae^{kx}$.

$$ak^2 + bk + c = 0$$
, Auxiliary equation

real

$$k = a, b$$

imaginary

$$k = \pm \beta i$$

complex $k = \alpha \pm \beta i$

$$y = Ae^{ax} + Be^{bx},$$

$$y = Ae^{aa} + Be^{aa}$$
,

$$y = A\sin(\beta x) + B\cos(\beta x),$$

$$y = e^{\alpha x} [A \sin(\beta x) + B \cos(\beta x)].$$

A,B from

ICS/BCS